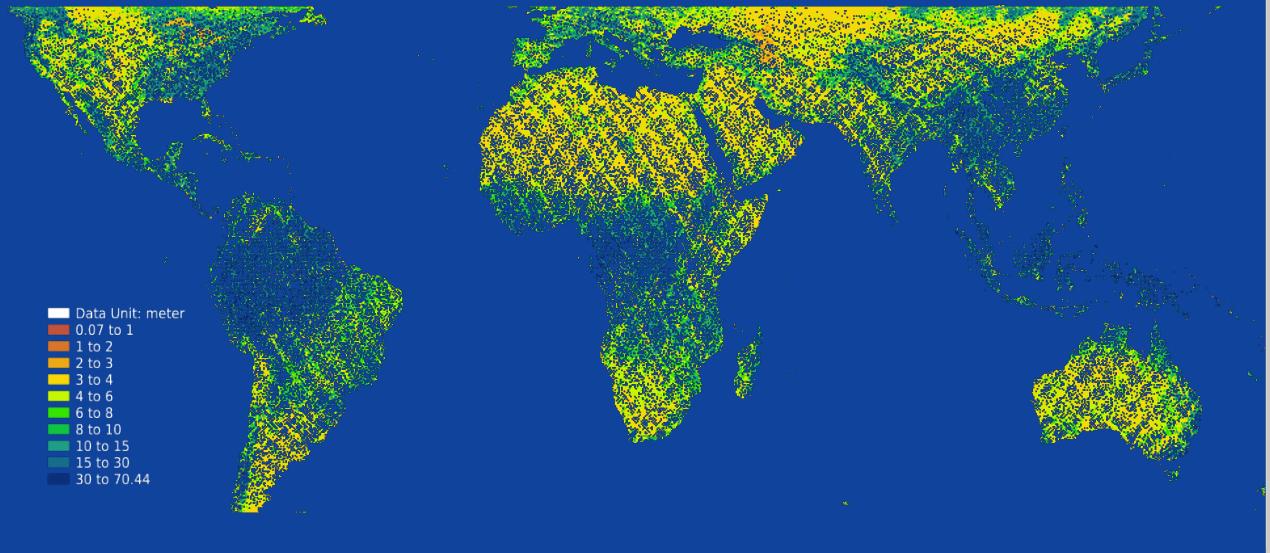


# Explore NASA GEDI Aboveground Biomass Datasets, Services, and Tools at ORNL DAAC

Rupesh Shrestha  
ORNL DAAC Staff Scientist

Bruce E. Wilson  
ORNL DAAC Manager



The Oak Ridge National Laboratory Distributed Active Archive Center for Biogeochemical Dynamics operates under an interagency agreement between NASA and the U.S. Department of Energy



**U.S. DEPARTMENT OF  
ENERGY**



# Outline

- Introduction (Bruce)
  - ORNL DAAC
  - GEDI mission and Data
  - Available services for GEDI data
    - Earthdata Cloud
- Examples of accessing and using GEDI data (Rupesh)
  - GEDI Level 3, Level 4A, and Level 4B data overview
  - Using S3 Direct Access Tutorial
  - Subsetting GEDI L4A using Harmony

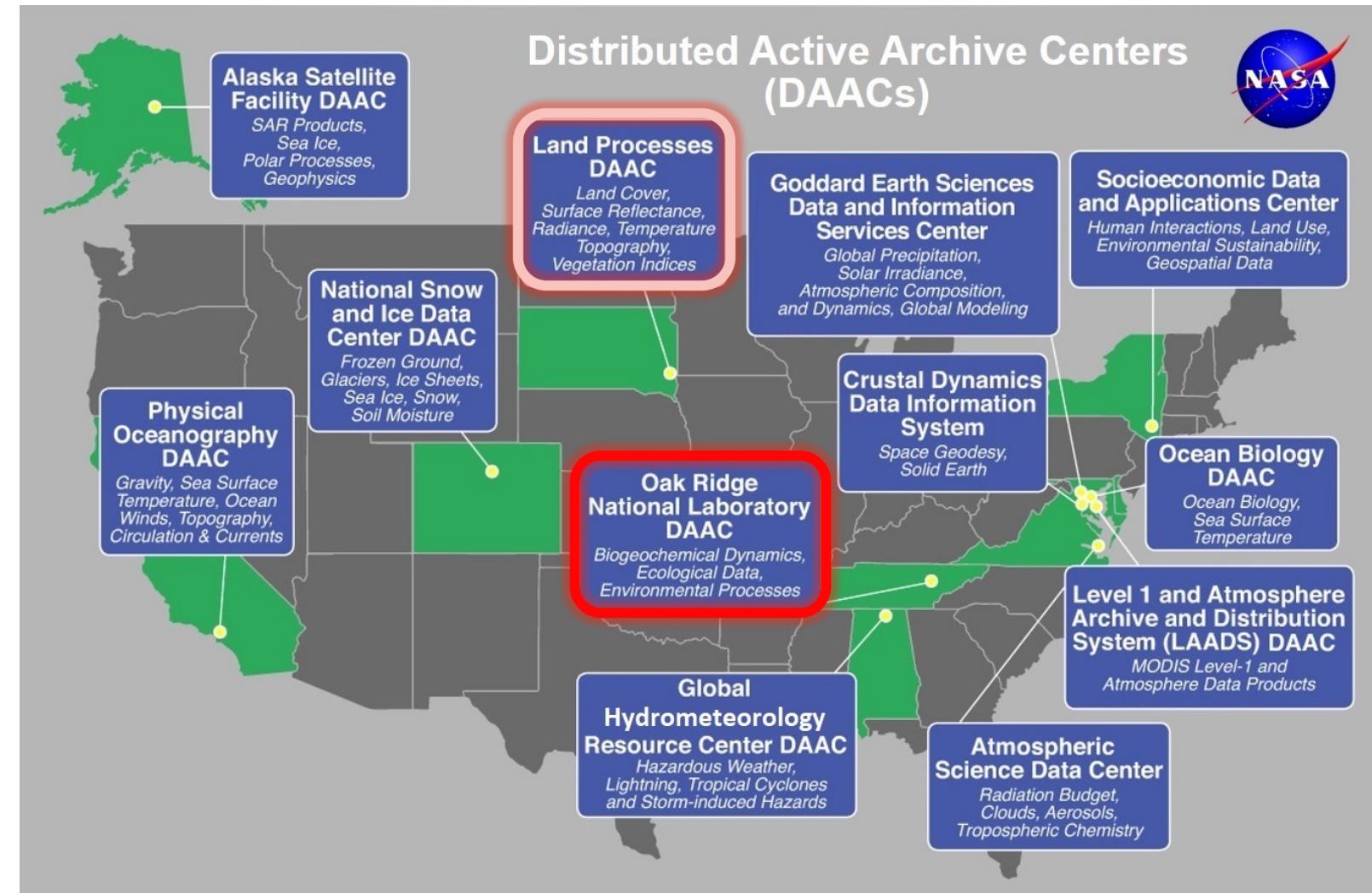


# Earth Observing System Data and Information System (EOSDIS) Distributed Active Archive Centers (DAACs)

Distributed Active Archive Centers (DAACs) are collocated with centers of science discipline expertise, archive and distribute NASA Earth Science data products.

<https://lpdaac.usgs.gov>

<https://daac.ornl.gov>



# The Oak Ridge National Laboratory Distributed Active Archive Center for Biogeochemical Dynamics

The ORNL DAAC archives and distributes terrestrial ecology data, particularly data from field and airborne campaigns.

<https://daac.ornl.gov>



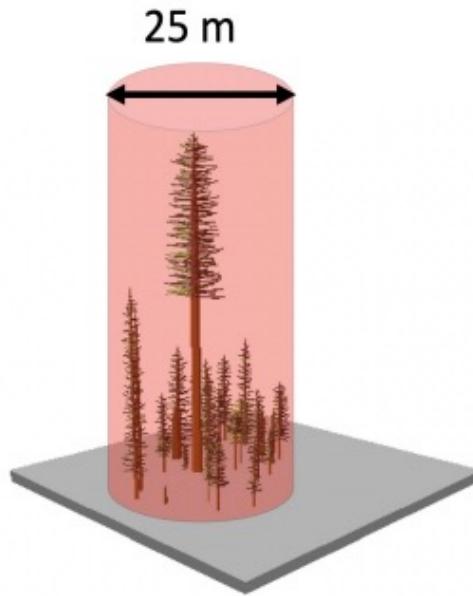
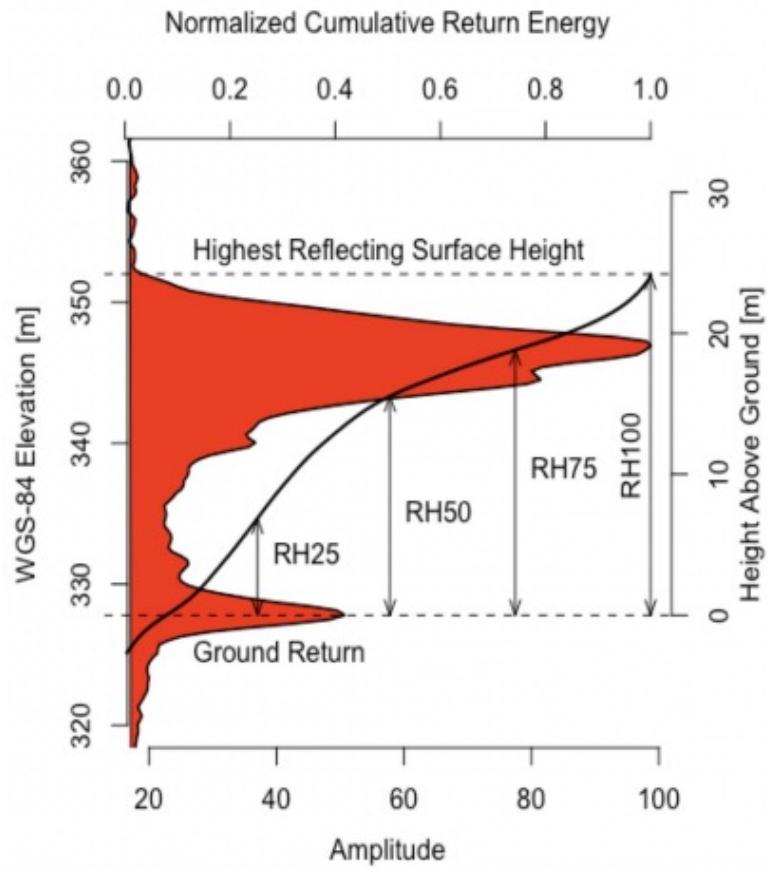
# Global Ecosystem Dynamics Investigation



- High resolution laser ranging of the Earth's forests & topography
- Operating on the ISS since April 2019
- 3 lasers, full waveform in NIR
- 8 tracks with 25 m footprints,  
600 m between tracks
- Coverage between +/- 52° latitude  
(ISS orbit)
- <https://gedi.umd.edu/>  
<https://daac.ornl.gov/gedi> (ORNL DAAC GEDI data)  
<https://bit.ly/2YgIDiE> (GEDI Overview at LP DAAC)  
<https://youtu.be/xsoMcw4upjs> (Meet GEDI! video by LP DAAC)

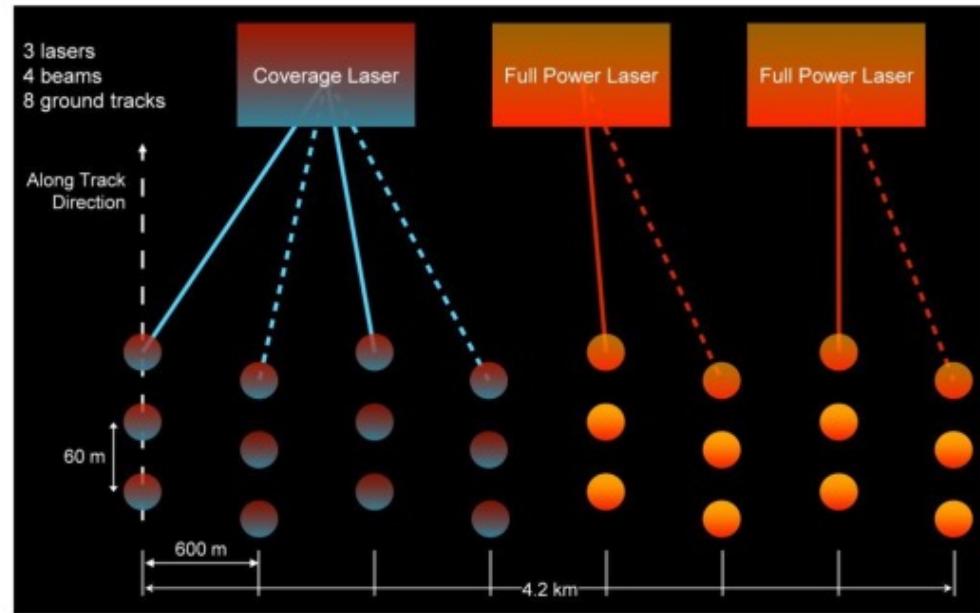


# GEDI is full waveform lidar

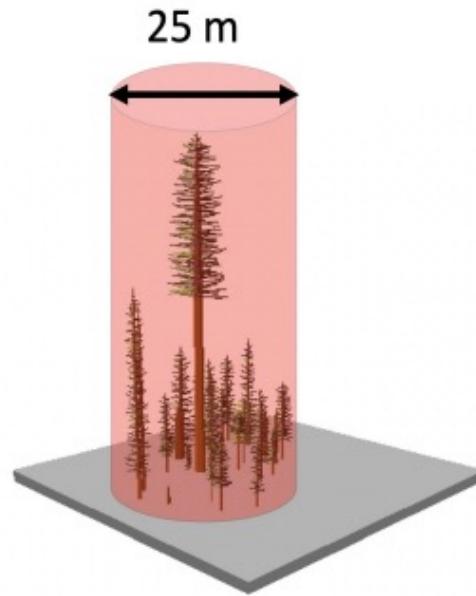
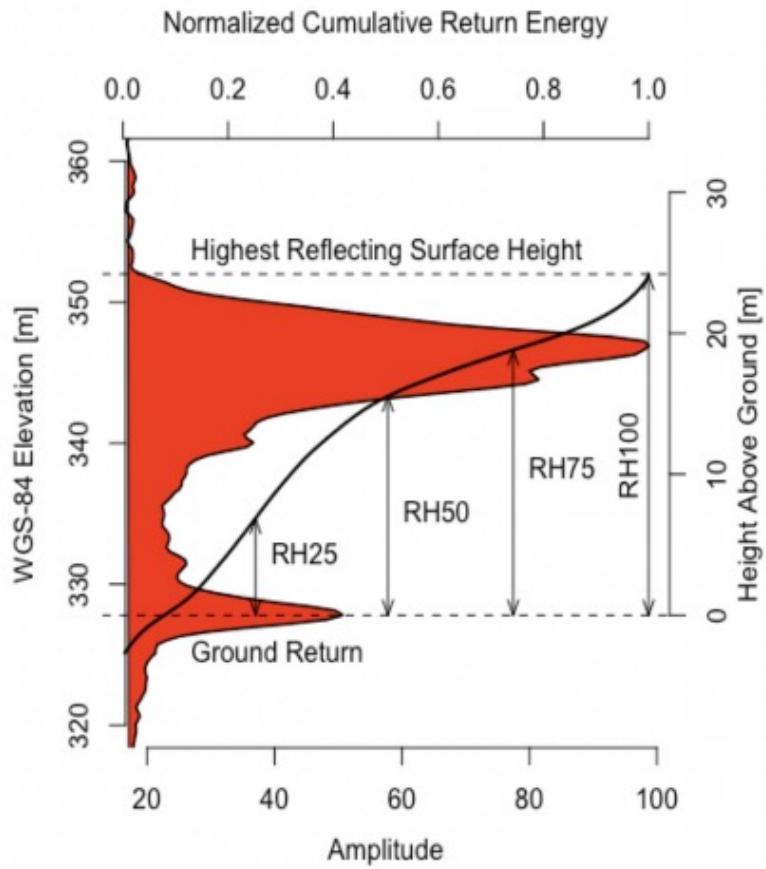


Level 1: Geolocated Waveform

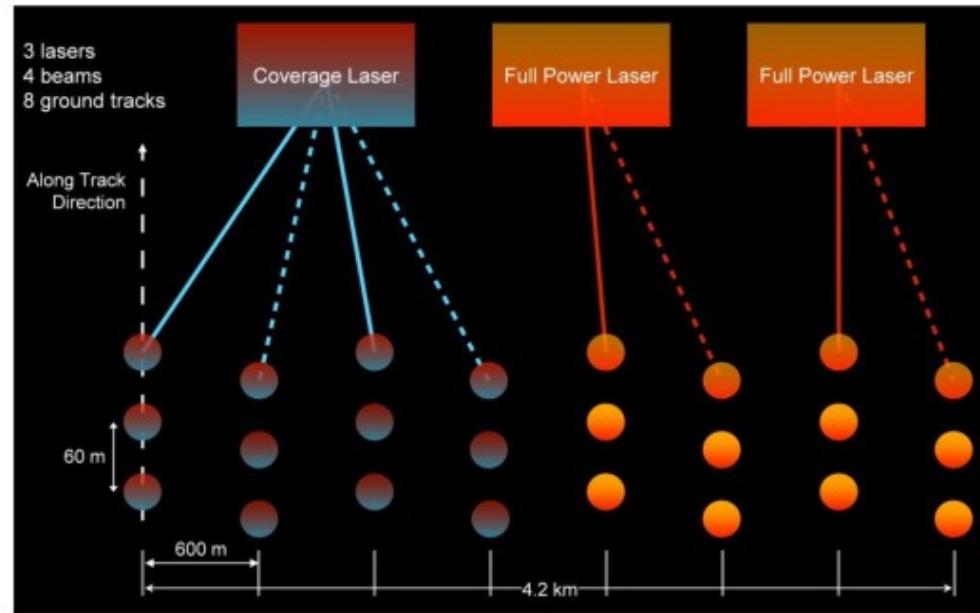
**GEDI's sole observable is the lidar waveform which provides ground elevation, canopy height, cover and various profiles and metrics**



# GEDI is full waveform lidar

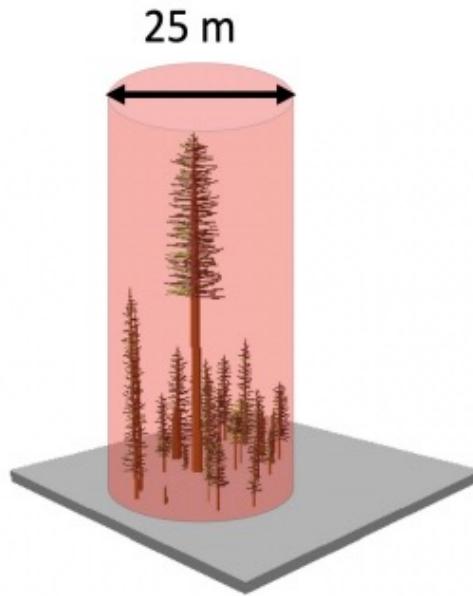
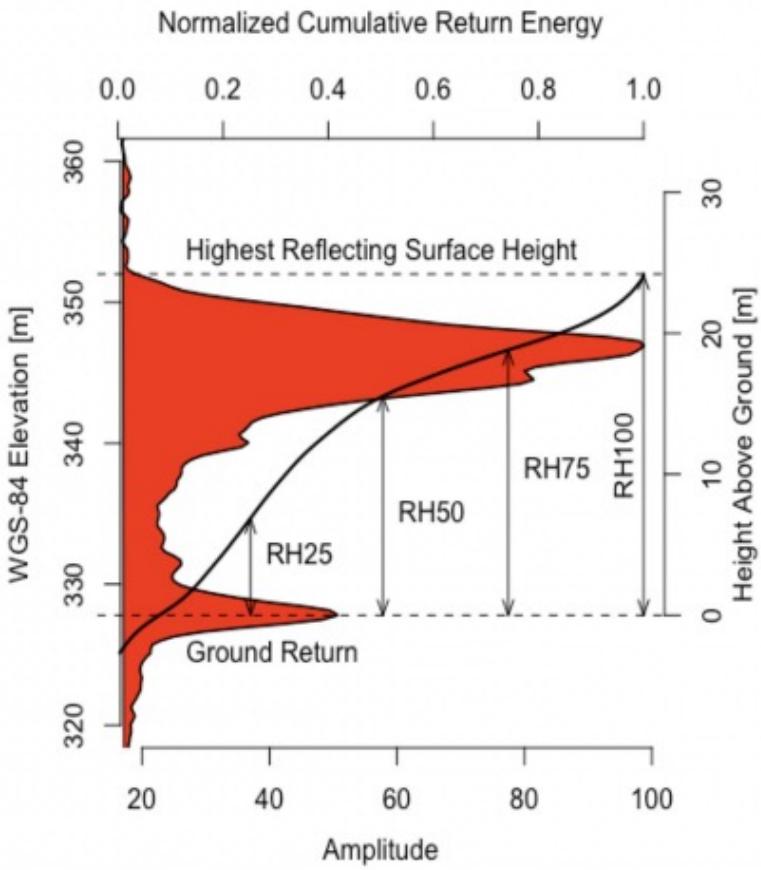


**GEDI's sole observable is the lidar waveform which provides ground elevation, canopy height, cover and various profiles and metrics**

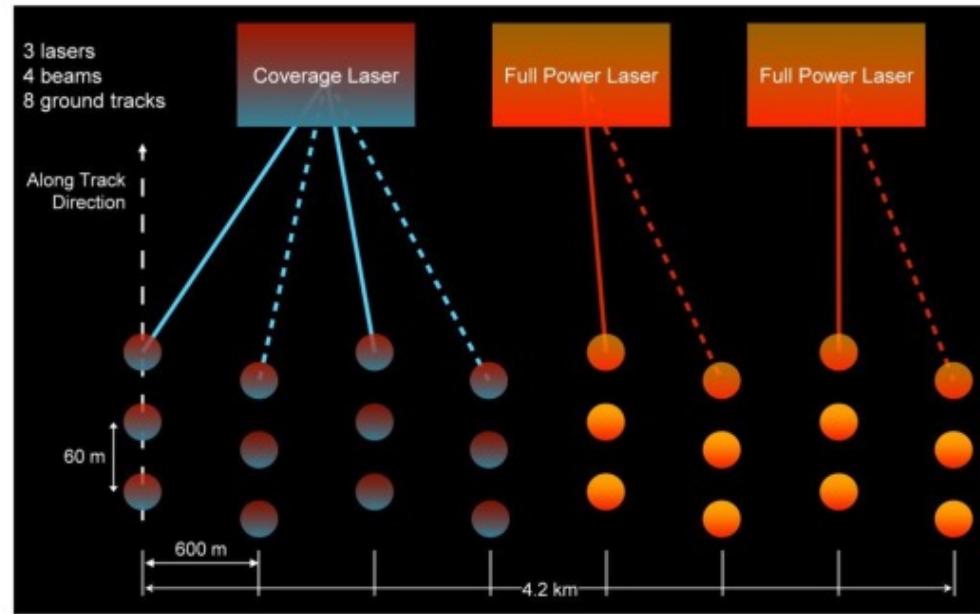


Level 1: Geolocated Waveform  
Level 2: Footprint level metrics

# GEDI is full waveform lidar

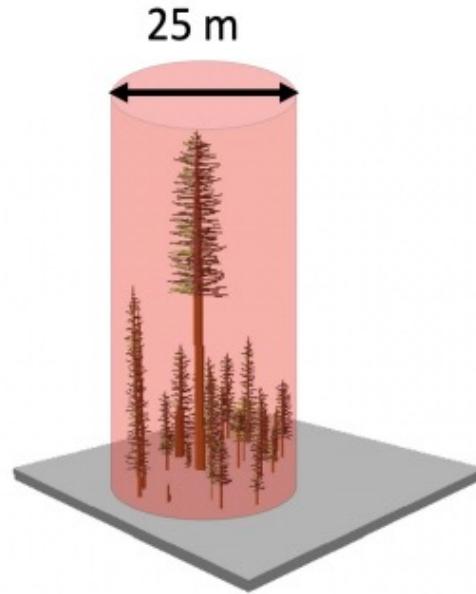
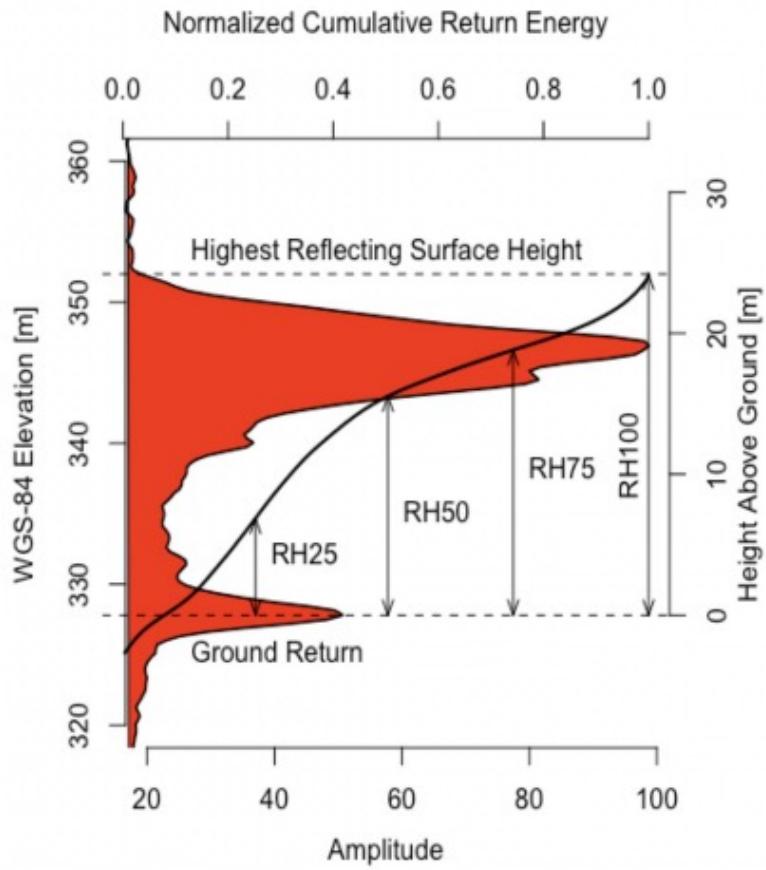


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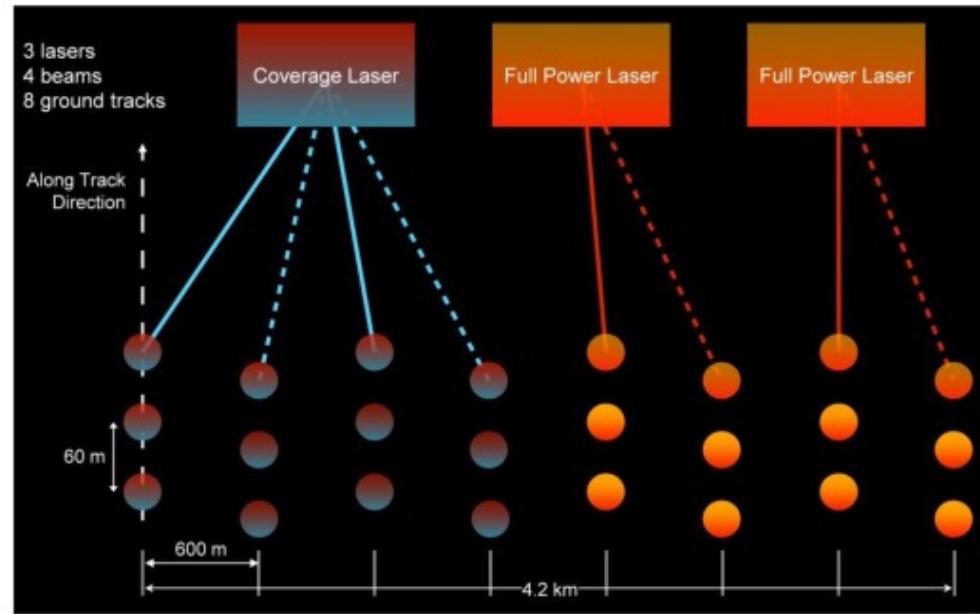


Level 1: Geolocated Waveform  
Level 2: Footprint level metrics  
Level 3: Gridded metrics

# GEDI is full waveform lidar

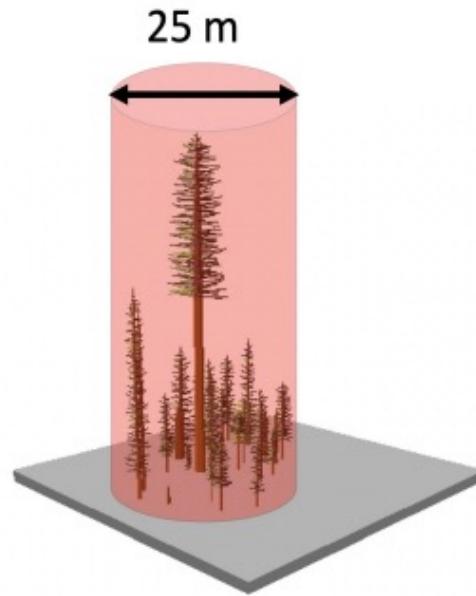
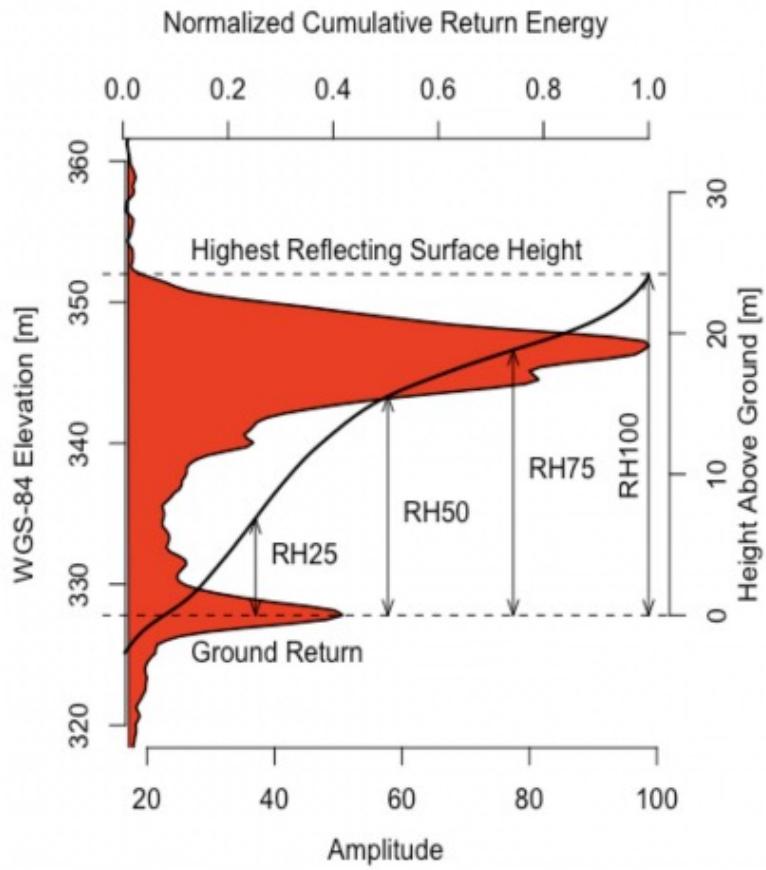


**GEDI's sole observable is the lidar waveform which provides ground elevation, canopy height, cover and various profiles and metrics**

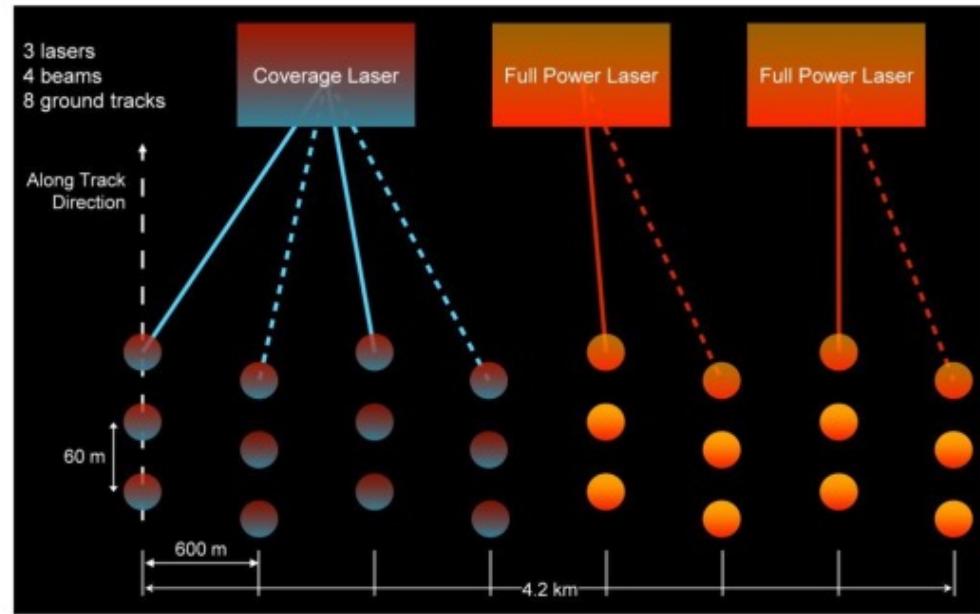


- Level 1: Geolocated Waveform
- Level 2: Footprint level metrics
- Level 3: Gridded metrics
- Level 4A: Footprint level aboveground biomass

# GEDI is full waveform lidar



**GEDI's sole observable is the lidar waveform which provides ground elevation, canopy height, cover and various profiles and metrics**



- Level 1: Geolocated Waveform
- Level 2: Footprint level metrics
- Level 3: Gridded metrics
- Level 4A: Footprint level aboveground biomass
- Level 4B: Gridded aboveground biomass

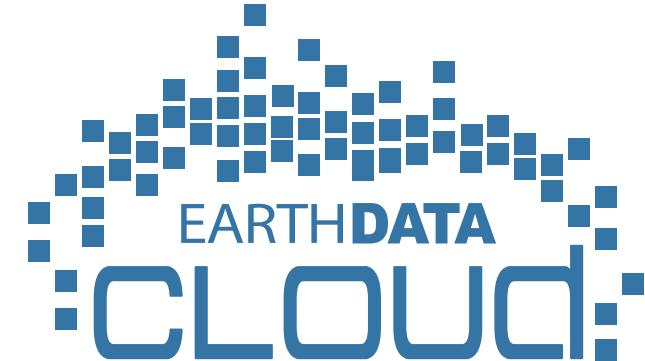
# GEDI data is available through LP and ORNL DAACs

Product	Description	Archive DAAC
Level 1B	Geolocated waveform	LP DAAC
Level 2A	Elevation and height metrics	
Level 2B	Canopy cover and vertical profile metrics	
Level 3	Gridded land surface metrics	ORNL DAAC
Level 4A	Footprint level aboveground biomass density	
Level 4B	Gridded aboveground biomass density	



# Earthdata Cloud: Further enabling Open Science

- NASA Earth Science Data and Services in a commercial (AWS) cloud environment
- A work in progress
- Tools for cross-DAAC access to data
  - Harmony: <https://harmony.earthdata.nasa.gov/>
  - OPeNDAP Hyrax: <https://opendap.earthdata.nasa.gov/>
  - Direct (in-region) S3 Access



## GEDI L4B Gridded Aboveground Biomass Density, Version 2

10 Granules 2019-04-18 to 2021-08-04

 Earthdata Cloud

This Global Ecosystem Dynamics Investigation (GEDI) L4B product provides 1 km x 1 km (1 km, hereafter) estimates of mean aboveground biomass ...



### Cloud Access

Available for access in-region with AWS Cloud

#### Region

us-west-2

#### Bucket/Object Prefix

s3://ornl-cumulus-prod-protected/gedi/GEDI\_L4B\_Gridded\_Biomass/

#### AWS S3 Credentials

[Get AWS S3 Credentials](#) 

[Documentation](#) 



# Higher level GEDI data is available in multiple tools

Tool/Service	L3	L4A	L4B
<a href="#">Earthdata Cloud S3 access</a>	✓	✓	✓
<a href="#">Earthdata Cloud OPeNDAP</a>		✓	
<a href="#">Earthdata Search</a> (Discovery)	✓	✓	✓
<a href="#">Common Metadata Repository API</a> (Discovery)	✓	✓	✓
<a href="#">WorldView</a> and <a href="#">Global Imagery Browse Services</a> (GIBS) API	✓		soon
<a href="#">HTTPS file download</a>	✓	✓	✓
<a href="#">Harmony API for subsetting</a>		✓	
<a href="#">Open Geospatial Consortium Web Mapping Service (OGC WMS)</a>	✓		✓
<a href="#">Open Geospatial Consortium Web Coverage Service (OGC WCS)</a>	✓		✓
<a href="#">ORNL DAAC's Spatial Data Access Tool (SDAT)</a>	✓		✓
<a href="#">ORNL DAAC's Shopping Cart</a>	✓	✓	✓
<a href="#">ORNL DAAC's Terrestrial Ecology Subsetting &amp; Visualization Services</a>	✓	✓	✓
<a href="#">ORNL DAAC's Quick Order</a>	✓		✓



Earthdata login (<https://urs.earthdata.nasa.gov>) required for all data access

L3 – Gridded (1 km) land surface metrics  
L4A – Footprint level aboveground biomass  
L4B – Gridded (1 km) aboveground biomass

# Higher level GEDI data is available in multiple tools

Tool/Service	L3	L4A	L4B
<a href="#"><u>Earthdata Cloud S3 access</u></a> ←	✓	✓	✓
<a href="#"><u>Earthdata Cloud OPeNDAP</u></a>		✓	
<a href="#"><u>Earthdata Search</u> (Discovery)</a>	✓	✓	✓
<a href="#"><u>Common Metadata Repository API</u> (Discovery)</a>	✓	✓	✓
<a href="#"><u>Worldview</u></a> and <a href="#"><u>Global Imagery Browse Services (GIBS)</u></a> API	✓		soon
<a href="#"><u>HTTPS file download</u></a>	✓	✓	✓
<a href="#"><u>Harmony API for subsetting</u></a> ←		✓	
<a href="#"><u>Open Geospatial Consortium Web Mapping Service (OGC WMS)</u></a>	✓		✓
<a href="#"><u>Open Geospatial Consortium Web Coverage Service (OGC WCS)</u></a>	✓		✓
<a href="#"><u>ORNL DAAC's Spatial Data Access Tool (SDAT)</u></a>	✓		✓
<a href="#"><u>ORNL DAAC's Shopping Cart</u></a>	✓	✓	✓
<a href="#"><u>ORNL DAAC's Terrestrial Ecology Subsetting &amp; Visualization Services</u></a>	✓	✓	✓
<a href="#"><u>ORNL DAAC's Quick Order</u></a>	✓		✓



Earthdata login (<https://urs.earthdata.nasa.gov>) required for all data access

L3 – Gridded (1 km) land surface metrics  
L4A – Footprint level aboveground biomass  
L4B – Gridded (1 km) aboveground biomass

# Outline

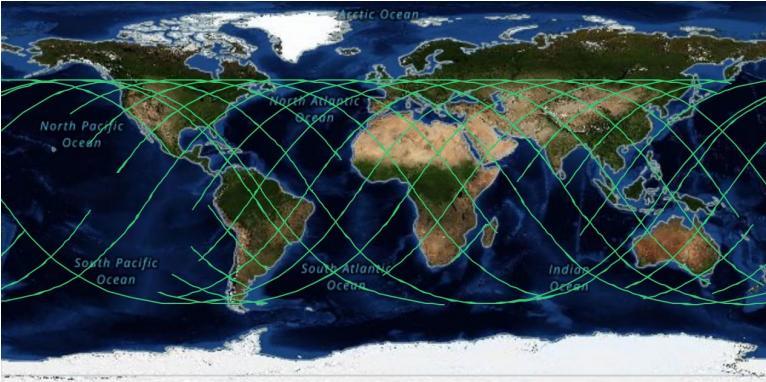
- Overview of GEDI Level 3, Level 4A and Level 4B
- Examples of accessing and using GEDI data
  - Using S3 Direct Access (Jupyter Notebook)
  - Using NASA Harmony API
    - Jupyter Notebook
    - Retrieving GEDI L4A subsets using NASA Earthdata Search



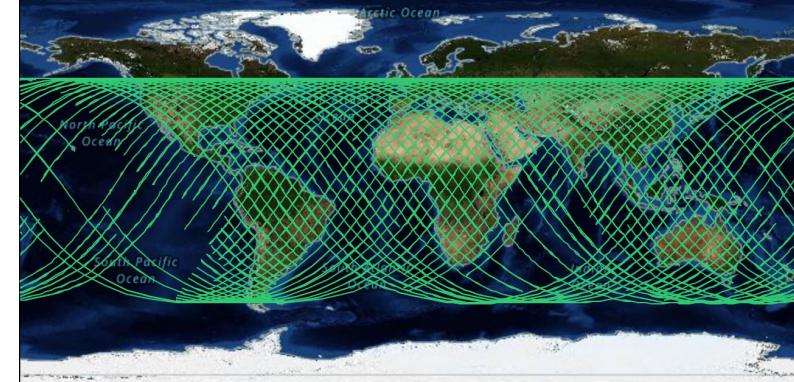
# GEDI Orbits, Footprints, Grids



One GEDI orbit (orbit number 06577) (2020-02-09 15:13:21 to 16:46:15) => 4 L4A granules



GEDI orbits over one-day (2020-02-09) => 58 L4A granules



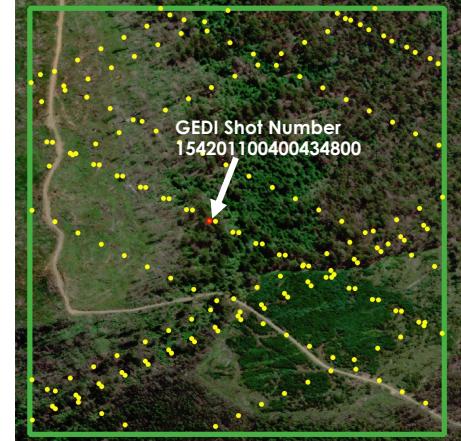
GEDI orbits over period of one-week (2020-02-09 to 2020-02-15) => 397 L4A granules



Green box = 1km x 1km EASEGRID in Tasmania, Australia. This is area where Australia's tallest trees are located



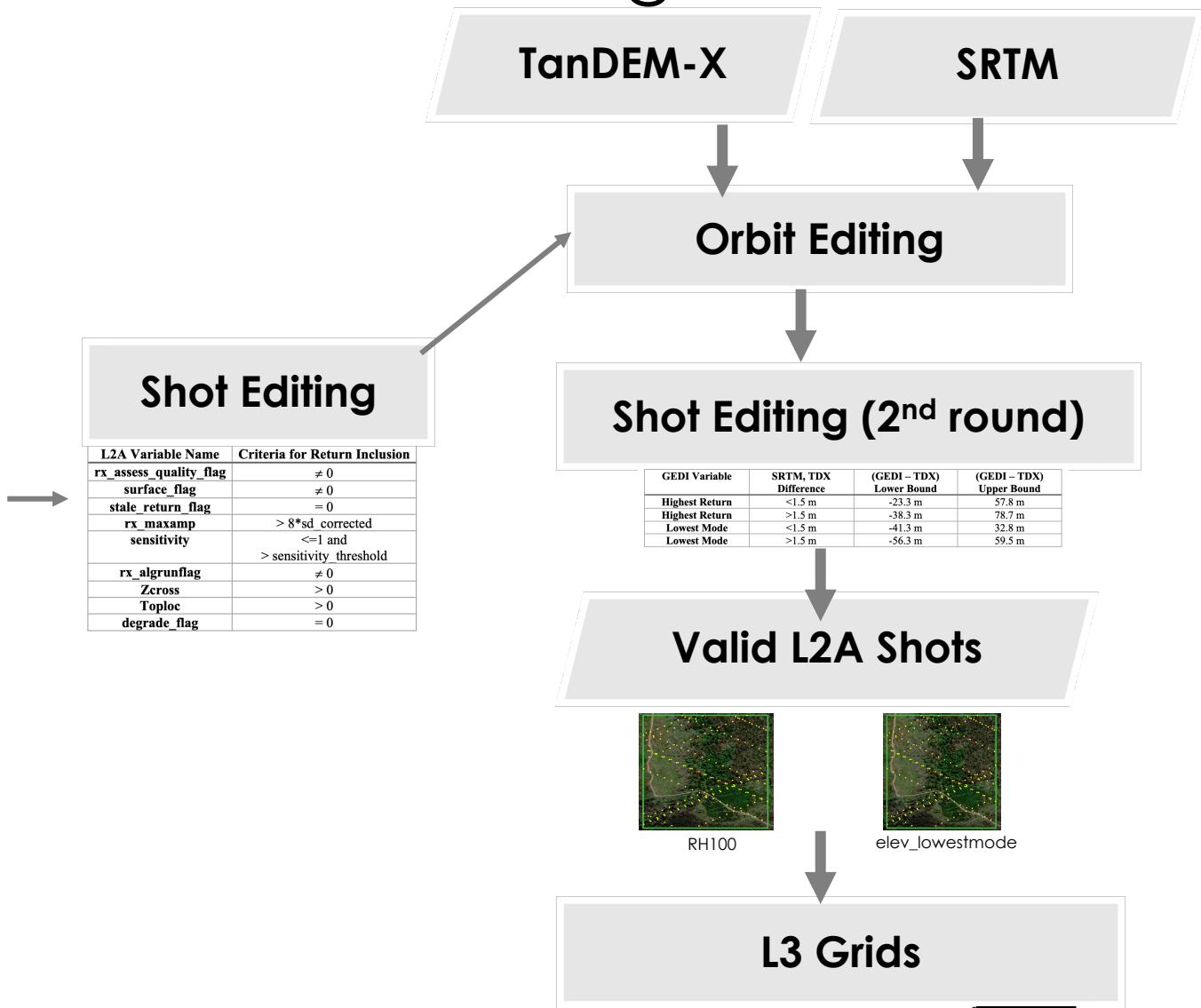
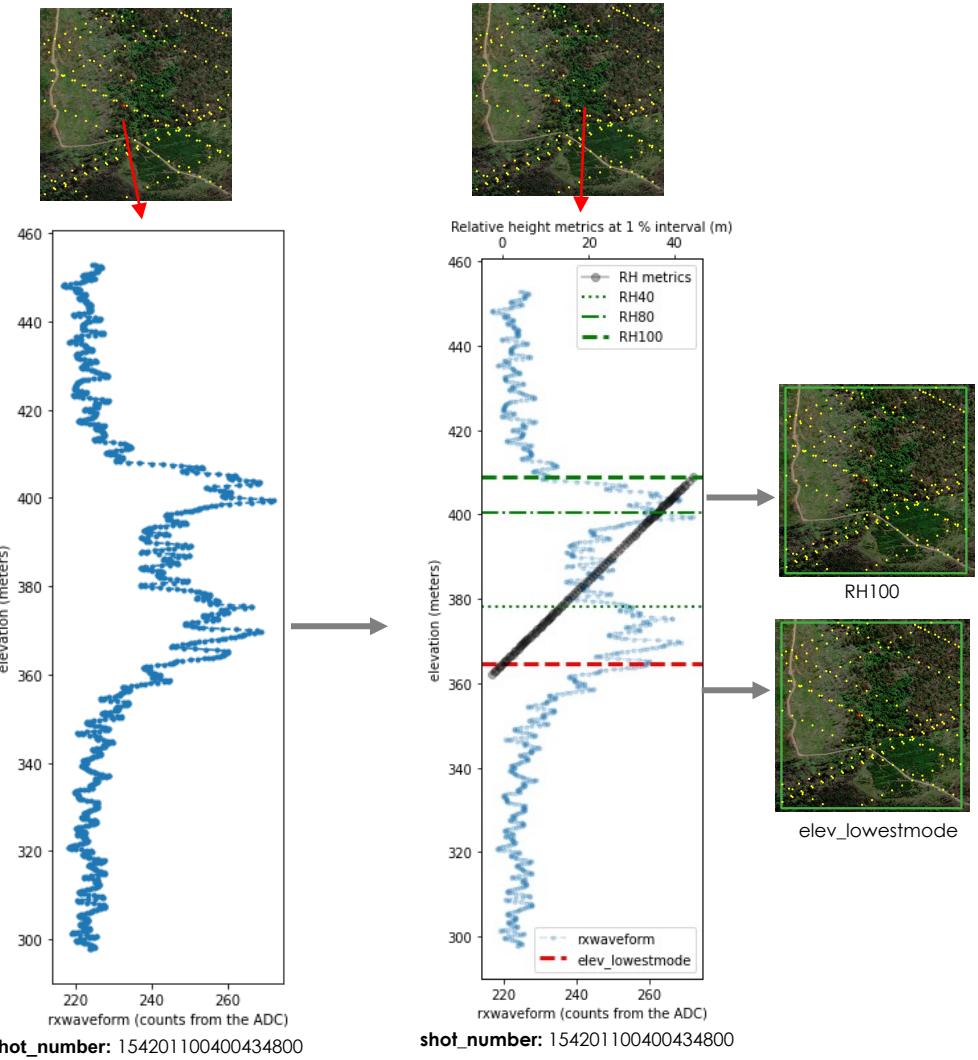
10 GEDI sub-orbits passes over the 1km x 1km grid between 2019-04-17 to 2022-01-20



Yellow dots = GEDI shots within 1km x 1km EASEGRID (193 GEDI shots)



# GEDI L3 Gridded Land Surface Metrics - Algorithm

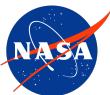
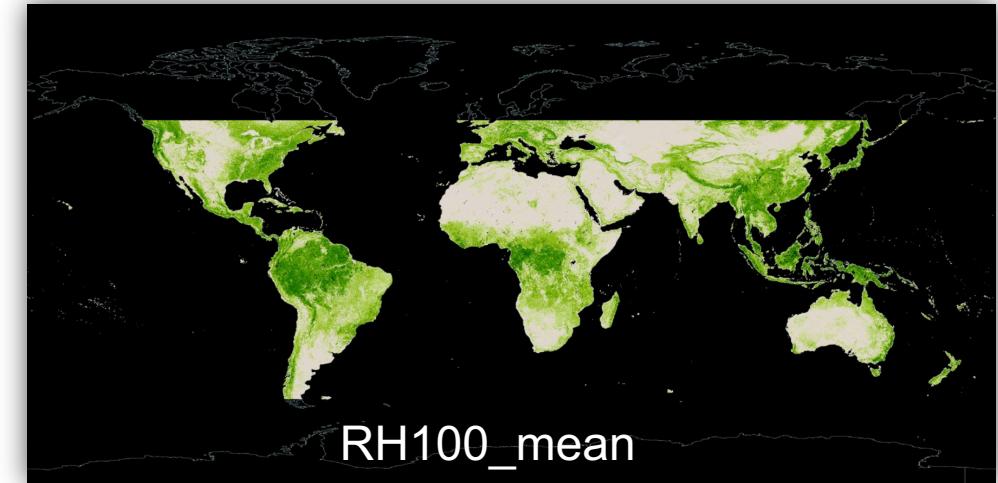
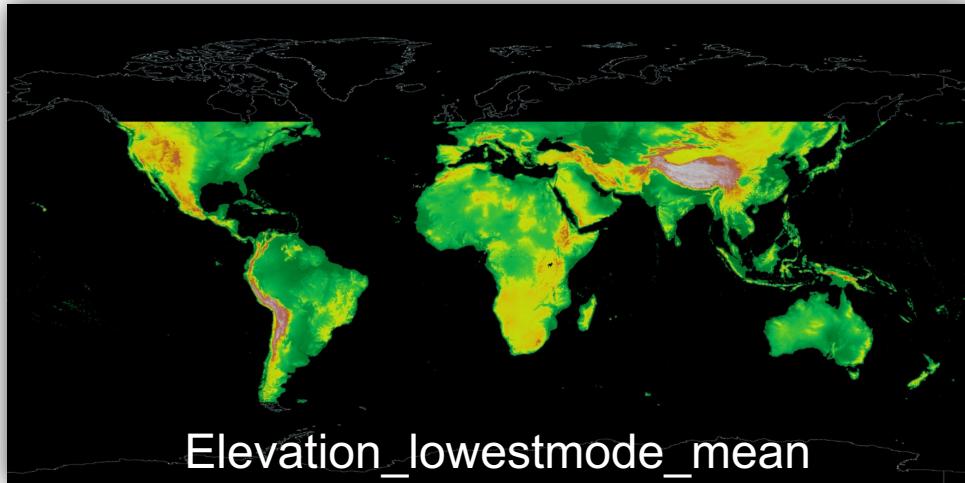


$$\bar{x} = \frac{1}{n} \sum x_i$$

$$\sigma = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n}}$$

# GEDI L3 Gridded Land Surface Metrics - Variables

Layers	Description
Counts	Number of valid laser footprints
Elevation_lowestmode_mean	Ground elevation approximated by the mean elevation of the lowest mode of valid footprints
Elevation_lowestmode_stddev	Standard deviation of the elevation of lowest mode of valid footprints
RH100_mean	Canopy height above ground in each grid cell characterized by the mean Relative Height Metrics 100 values of valid footprints
RH100_stddev	Standard deviation of RH100 for all valid footprints

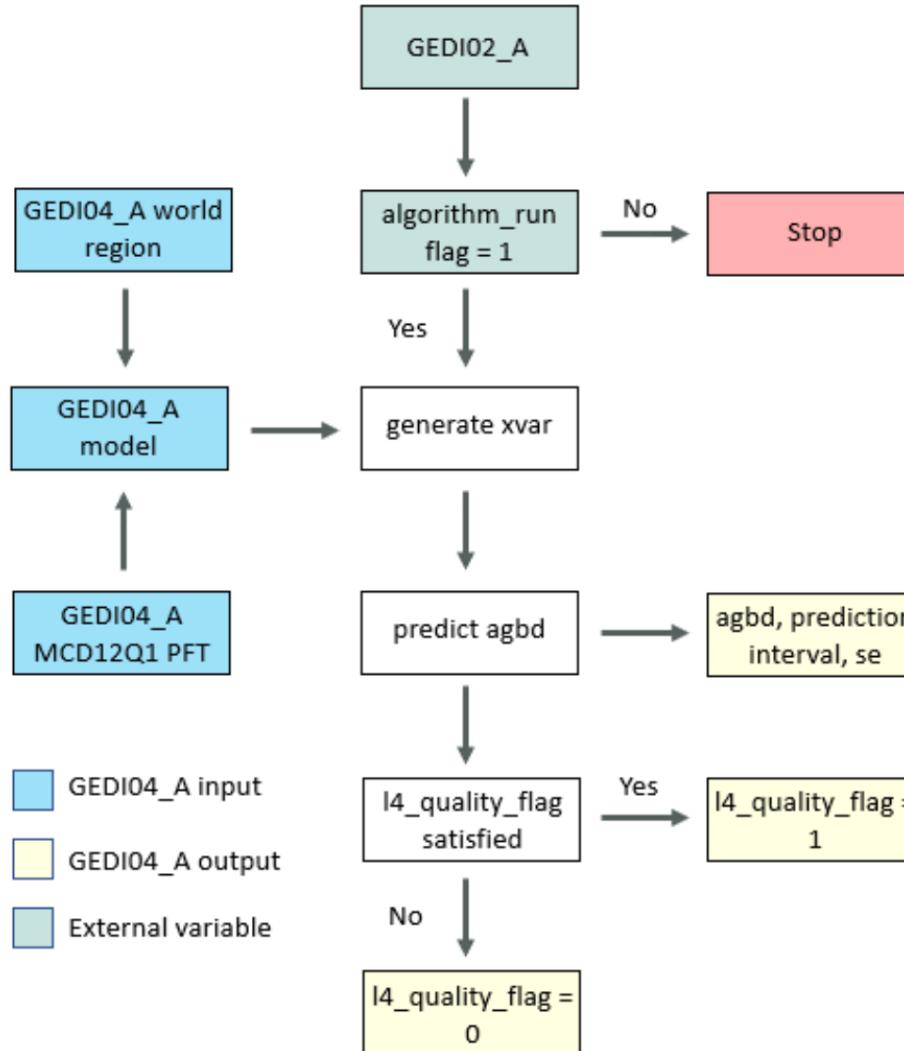
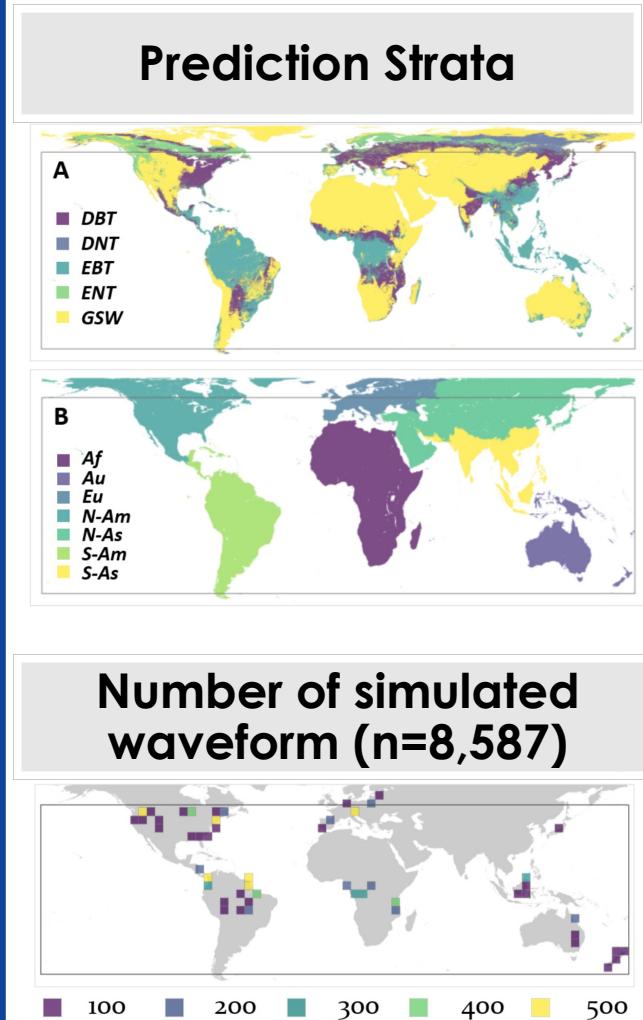


# GEDI L3 Gridded Land Surface Metrics

- **Data Format:** Cloud-optimized GeoTIFFs
- **Grid System:** 1km EASEGrid 2.0
- **Spatial Extent:** -52N to 52 degrees latitude
- **Temporal Extent:** Mission Weeks 19-96, 19-122, 19-138, 19-162
- **Variables:** 5 layers



# GEDI L4A Footprint Level AGBD - Algorithm



**Algorithm Setting Groups**

Algorithm	Rx_processing Subgroup	Smooth width	Smoothwidth_zcross	Front_threshold	Back_threshold
1	a1	6.5	6.5	3σ	6σ
2	a2	6.5	3.5	3σ	3σ
3	a3	6.5	3.5	3σ	6σ
4	a4	6.5	6.5	6σ	6σ
5	a5	6.5	3.5	3σ	2σ
6	a6	6.5	3.5	3σ	4σ



# GEDI L4A Footprint Level AGBD

- **Data Format:** HDF5
- **Spatial Extent:** -52N to 52 degrees latitude
- **Temporal Extent:** Mission Weeks 19-138
- **Variables:** 183 variables within each of 8 beams, Ancillary and metadata information



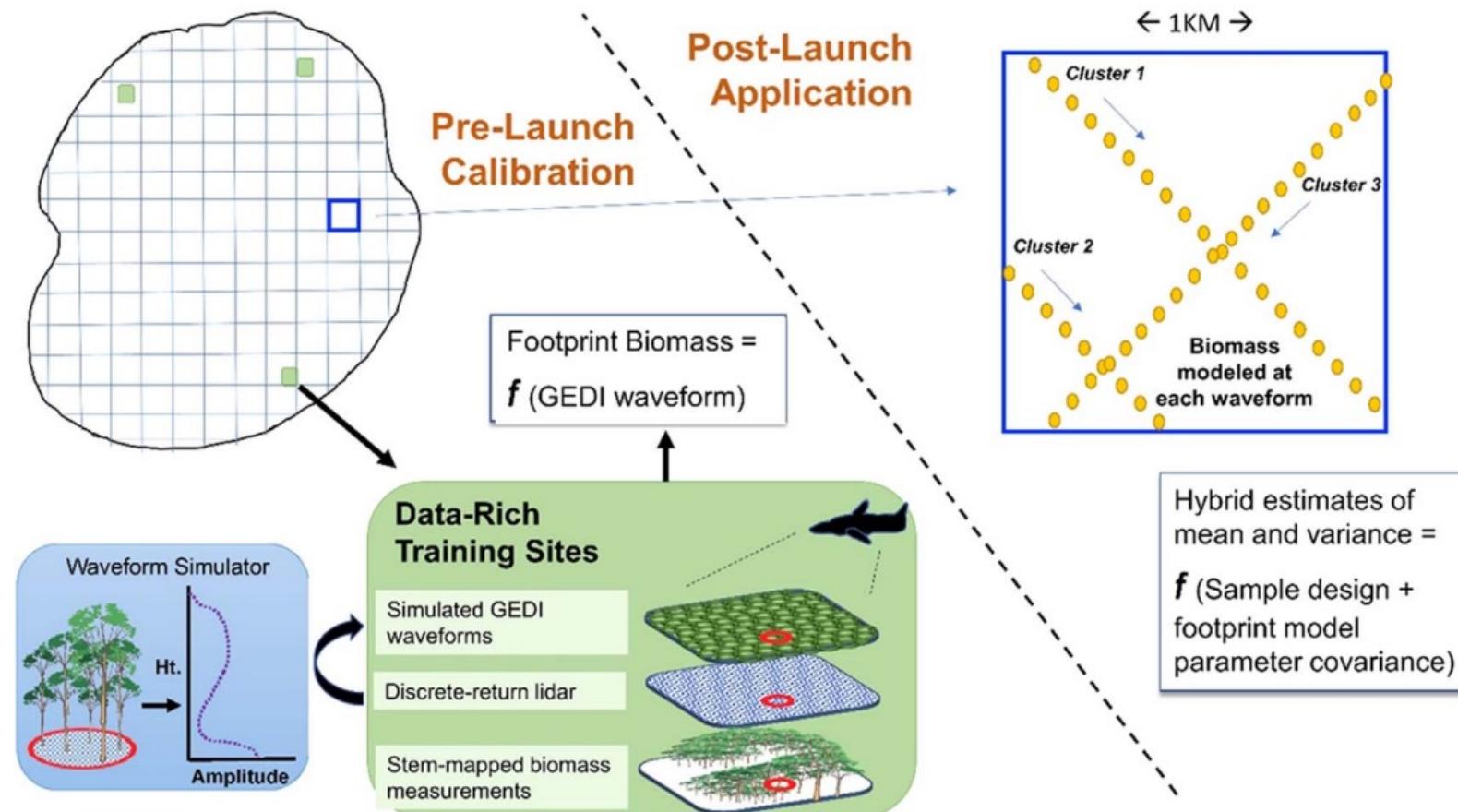
# GEDI L4A Footprint Level AGBD - Variables

variable	description	units	source
agbd	Aboveground biomass density (Mg / ha)	Mg / ha	
agbd_pi_lower	Lower prediction interval (see alpha attribute for the level)	Mg / ha	
agbd_pi_upper	Upper prediction interval (see alpha attribute for the level)	Mg / ha	
agbd_se	Aboveground biomass density (Mg / ha) prediction standard error	Mg / ha	
agbd_t	Model prediction in fit units	-	
agbd_t_se	Model prediction standard error in fit units (needed for calculation of custom prediction intervals)	-	
algorithm_run_flag	The L4A algorithm is run if this flag is set to 1. This flag selects data which have sufficient waveform fidelity for AGBD estimation.	-	
l2_quality_flag	Flag identifying the most useful L2 data for biomass predictions	-	
l4_quality_flag	Flag simplifying selection of most useful biomass predictions	-	
predict_stratum	Character ID of the prediction stratum name for the 1 km cell	-	
predictor_limit_flag	Predictor value is outside the bounds of the training data (0=In bounds; 1=Lower bound; 2=Upper bound)	-	
response_limit_flag	Prediction value is outside the bounds of the training data (0=In bounds; 1=Lower bound; 2=Upper bound)	-	
xvar	Predictor variables (offset and transformation have been applied)	-	



**User Guide:** [https://daac.ornl.gov/GEDI/guides/GEDI\\_L4A\\_AGB\\_Density\\_V2\\_1.html](https://daac.ornl.gov/GEDI/guides/GEDI_L4A_AGB_Density_V2_1.html)  
**Tutorial:** [https://github.com/ornldaac/gedi\\_tutorials/blob/main/3\\_gedi\\_l4a\\_exploring\\_data.ipynb](https://github.com/ornldaac/gedi_tutorials/blob/main/3_gedi_l4a_exploring_data.ipynb)

# GEDI L4B Gridded AGBD - Algorithm



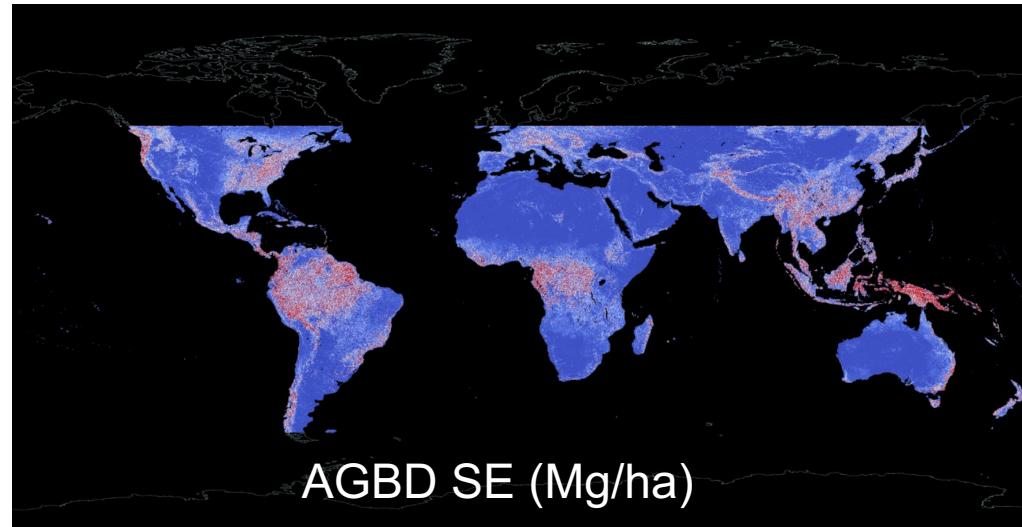
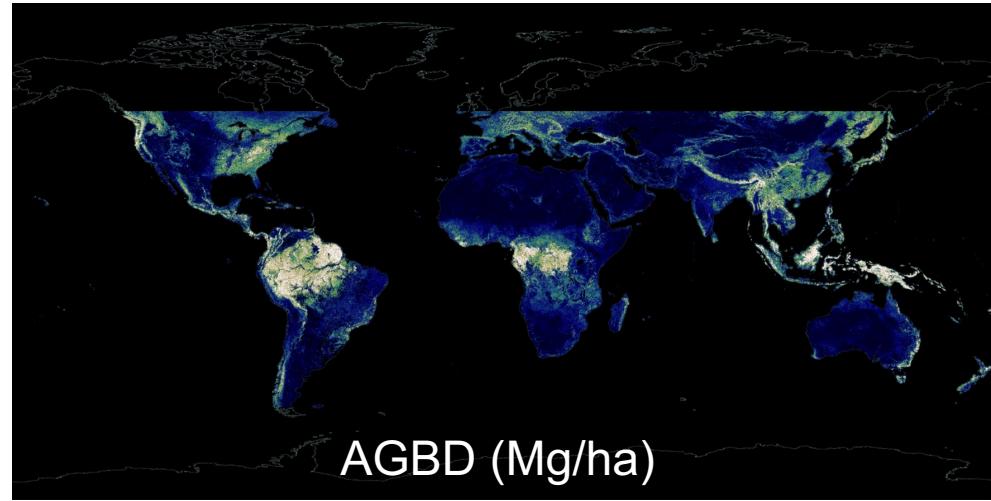
**Source:** Patterson et al. 2019 <https://doi.org/10.1088/1748-9326/ab18df>



**ATBD:** [https://daac.ornl.gov/daacdata/gedi/GEDI\\_L4B\\_Gridded\\_Biomass/comp/GEDI\\_L4B\\_ATBD\\_v1.0.pdf](https://daac.ornl.gov/daacdata/gedi/GEDI_L4B_Gridded_Biomass/comp/GEDI_L4B_ATBD_v1.0.pdf)  
**Manuscript:** Dubayah et al. (in review) <https://eartharxiv.org/repository/object/3278/download/6566/>

# GEDI L4B Gridded AGBD

- **Data Format:** Cloud-optimized GeoTIFFs
- **Grid System:** 1km EASEGrid 2.0
- **Spatial Extent:** -52N to 52 degrees latitude
- **Temporal Extent:** Mission Weeks 19-138
- **Variables:** 10 layers



# GEDI L4 Gridded Land Surface Metrics - Variables

Layers	Description
Mean aboveground biomass density (MU)	Estimated mean AGBD, including forest and non-forest
Variance component 1 (V1)	Uncertainty in the estimate of mean biomass due to the field-to-GEDI model used in L4A
Variance component 2 (V2)	If Mode of Inference = 1, this is the uncertainty due to GEDI's sampling of the 1 km cell. If Mode of Inference = 2, this is uncertainty owing to the model predicting biomass
Mean aboveground biomass density standard error (SE)	Standard Error of the mean estimate, combining sampling and modeling uncertainty
Standard error as a fraction of the estimated mean AGBD (PE)	If >100%, the cell values are truncated to 100.
Number of clusters (NC)	Number of unique GEDI ground tracks with at least one high-quality waveform intersecting the grid cell
Number of samples (NS)	Total number of high-quality waveforms across all ground tracks within the grid cell
Quality flag (QF)	0=Outside the GEDI domain; 1=Land surface; 2=Land surface and meets GEDI mission L1 requirement (Percent standard error <20% or Standard Error < 20 Mg ha <sup>-1</sup> )
Prediction stratum (PS)	determined by plant functional type and continent. PS is associated with an L4A model parameter covariance matrix that contributes to the Model Error Variance
Mode of inference (MI)	Method used for a particular cell. Until mission completion, only those cells where hybrid inference is possible will be populated with a mean biomass value. 0=None applied; 1=Hybrid Model-Based; 2=Generalized Hierarchical Model-Based



# Learning Resources for GEDI Products

## Tutorials on GEDI Science Data Products

Author: ORNL DAAC

Date: August 26, 2021

Contact for the ORNL DAAC: uso@daac.ornl.gov

Keywords: lidar, GEDI, AGBD, aboveground biomass

### Overview

These tutorials demonstrate how to discover, access and use [GEDI science data products](#) archived at the ORNL DAAC.

### GEDI L4A Jupyter Notebooks

1. [Searching and downloading GEDI L4A dataset \(script\)](#)
2. [Subsetting GEDI L4A footprints \(script\)](#)
3. [Exploring GEDI L4A data structure](#)
4. [Accessing GEDI L4A dataset with NASA OPeNDAP in the Cloud \(script\)](#)
5. [Accessing GEDI L4A dataset with NASA Harmony API](#)
6. [Reproducing L4A AGBD estimates from GEDI L2A RH metrics](#)
7. [Apply correction to AGBD estimates for selected L4A shots, Version 2](#)

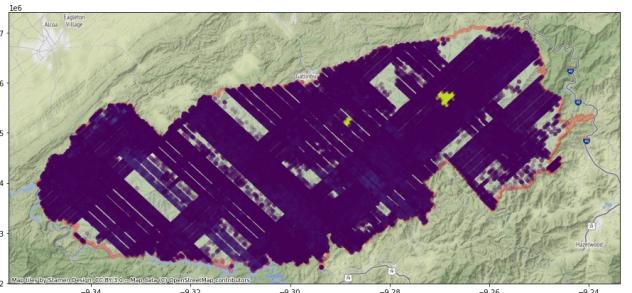
### GEDI L4B Jupyter Notebooks

1. [Access GEDI L4B Dataset with OGC Web Services](#)

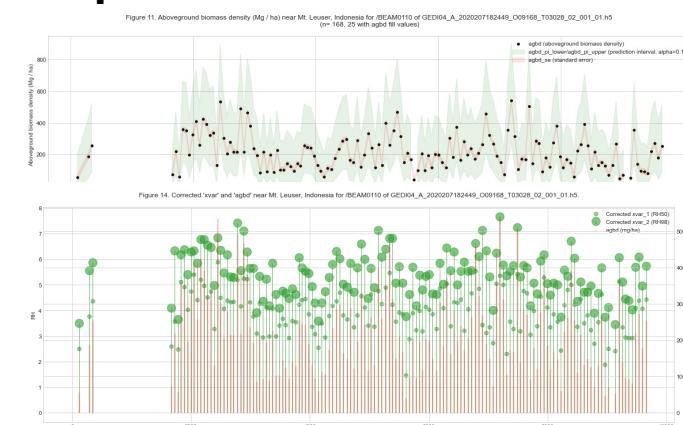
[https://github.com/ornldaac/gedi\\_tutorials/](https://github.com/ornldaac/gedi_tutorials/)

## Search and Download

### Subset



### Explore



### Python scripts

#### 1. gedi\_l4a\_search\_download.py

This script downloads GEDI L4A granules to a local directory based on GeoJSON polygon and start/end dates. First, set up NASA Earthdata Login authentication using a `.netrc` file. Please refer to the instructions here: <https://wiki.earthdata.nasa.gov/display/EL/How+To+Access+Data+With+cURL+And+Wget>.

usage

```
./gedi_l4a_search_download.py --doi <DOI> --date1 <start_date> --date2 <end_date> --poly <path_to_geojson_file> --outdir <
```

#### arguments

argument	description
--help	show help message and exit
--doi	dataset DOI e.g., 10.3334/ORNLDAC/2056 for GEDI L4A V2.1
--date1	start date in YYYY-MM-DD format
--date2	end date in YYYY-MM-DD format
--poly	path to a GeoJSON file defining area of interest
--outdir	path to the directory for saving downloaded h5 files

#### example usage

```
./gedi_l4a_search_download.py --date1 2019-12-15 --date2 2020-01-12 --doi 10.3334/ORNLDAC/2056 --poly ..//polygons/amapa.json
```

#### 2. gedi\_l4a\_subsets.py

This script subsets the downloaded GEDI L4A granules by a GeoJSON polygon file. The output files are in the H5 native format, with the option of converting to CSV or GeoJSON formats, and include the GEDI shots within the bounds of the polygon file.

usage

```
./gedi_l4a_subsets.py --poly <path_to_geojson_file> --indir <path_to_input_directory> --subdir <path_to_output_directory>
```

#### arguments

argument	description
--help	show help message and exit
--poly	path to a GeoJSON file defining area of interest
--indir	path to the directory with downloaded h5 files
--subdir	path to the directory for saving subset files
--csv	(optional) setting this creates additional output CSV subset file
--json	(optional) setting this creates additional output GeoJSON subset file

#### example usage

```
./gedi_l4a_subsets.py --poly ..//polygons/amapa.json --indir ..//full_orbits/ --subdir ..//subsets/ --csv
```

#### 3. gedi\_l4a\_hyrax.py

This script accesses the GEDI L4A dataset using NASA's OPeNDAP Hyrax. First, set up NASA Earthdata Login authentication using a `.netrc` file. Please refer to the instructions here: <https://wiki.earthdata.nasa.gov/display/EL/How+To+Access+Data+With+cURL+And+Wget>.

usage

```
./gedi_l4a_hyrax.py --doi <DOI> --date1 <start_date> --date2 <end_date> --poly <path_to_geojson_file> --beams <gedi_beams>
```



# Direct S3 Access GEDI L4A from the NASA Earthdata Cloud

GEDI L4A Datasets are available through NASA's Earthdata Cloud. NASA Earthdata on Cloud is always free and accessible via either HTTPS or direct [S3](#) bucket access. With direct S3 access, you can bring your "code to the data", making your processing faster and scalable. Direct S3 access to NASA Earthdata on Cloud is only available if your Amazon Web Services ([AWS](#)) instance is set up in the `us-west-2` region. If you are new to the Earthdata Cloud, these NASA Earthdata [primers](#) and [tutorials](#) are good resources to get you started.

To access a Jupyter Notebook on [AWS EC2](#) instance:

1. On the AWS EC2 instance, start Jupyter Notebook on port 8888 with the 'no-browser' parameter: `jupyter notebook --no-browser --port=8888`
2. On your local machine, forward port 8000 to the remote port 8888 : `ssh -i my-key-pair.pem -L 8000:localhost:8888 my-instance-user-name@my-instance-IPv6-address`
3. Now, the Jupyter Notebook can be accessed at `http://localhost:8000` in your local browser.

**Important:** [NASA Earthdata Login \(EDL\)](#) is required to obtain the S3 temporary credentials and direct access S3 objects bucket. First, set up NASA Earthdata Login authentication using a `.netrc` file. Please refer to the instructions here:

<https://wiki.earthdata.nasa.gov/display/EL/How+To+Access+Data+With+cURL+And+Wget>.

In this tutorial, we will retrieve GEDI L4A dataset from the Earthdata Cloud using direct S3 access.

[https://github.com/ornldaac/gedi\\_tutorials/blob/main/gedi\\_l4a\\_direct\\_s3\\_access.ipynb](https://github.com/ornldaac/gedi_tutorials/blob/main/gedi_l4a_direct_s3_access.ipynb)

```
In [1]: # import python modules
import requests
import s3fs
import h5py
import geopandas as gpd
import contextily as ctx
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from shapely.ops import orient
from IPython import display
from shapely.geometry import MultiPolygon, Polygon
from datetime import datetime
from multiprocessing import Pool
from posixpath import splitext
from os import path
import warnings
warnings.filterwarnings('ignore')
```

```
In [3]: # installing missing modules
!pip3 install --user s3fs
```

Requirement already satisfied: s3fs in ./local/lib/python3.8/site-packages (2022.5.0)

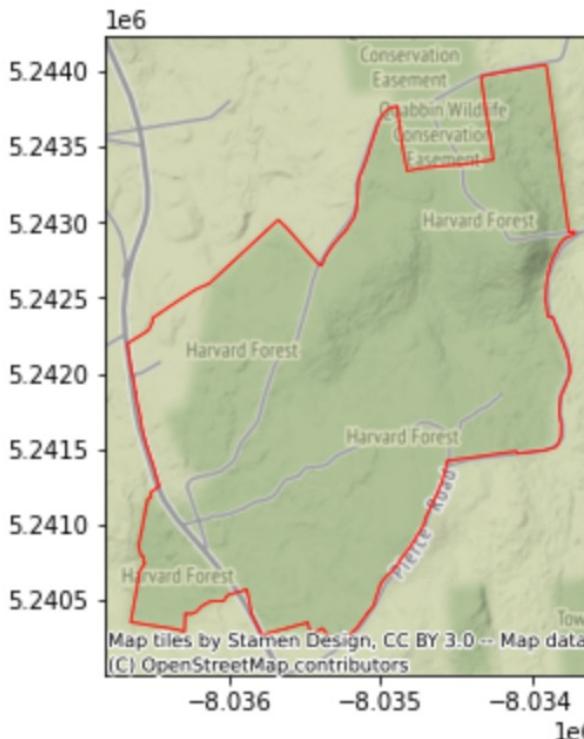
We will read an area within the [Harvard Forests](#) as a GeoJSON file. The Harvard Forests is one of the most extensively studied ecological sites in the United States.

If an area of interest is already defined as a polygon, the polygon file (geojson, shapefile, or kml) can be used to find overlapping GEDI L4A files. More details about this capability are described on [this page](#).

In [2]:

```
# read geojson
harvard = gpd.read_file("polygons/harvard.json")
# orient
harvard.geometry = harvard.geometry.apply(orient, args=(1,))

ax=harvard.to_crs(epsg=3857).plot(figsize=(5, 5), edgecolor='red', facecolor='none')
ctx.add_basemap(ax, zoom = 14)
```



Let's define a function that searches NASA's Common Metadata Repository ([CMR](#)) API to search and find the granules.

```
In [3]: def search_cmr(doi: str, geojson):
    # CMR API base url
    cmrurl='https://cmr.earthdata.nasa.gov/search/'
    # doi search to get concept_id
    doisearch = f'{cmrurl}collections.json?doi={doi}'
    concept_id = requests.get(doisearch).json()['feed'][0]['id']

    page_num = 1
    page_size = 2000 # CMR page size limit
    s3_arr = []

    while True:
        # defining parameters
        cmr_param = {
            "collection_concept_id": concept_id,
            "page_size": page_size,
            "page_num": page_num,
            "simplify-shapefile": 'true' # this is needed to bypass 5000 coordinates limit of CMR
        }

        granulesearch = f'{cmrurl}granules.json'
        r = requests.post(granulesearch, data=cmr_param, files=geojson)
        r.raise_for_status()
        granules = r.json()['feed']
        if granules:
            for g in granules:
                # Get s3 links
                for links in g['links']:
                    if links['href'].startswith('s3://'):
                        s3_arr.append(links['href'])

            page_num += 1
        else:
            break

    return s3_arr
```

```
In [4]: doi = '10.3334/ORNLDAAAC/2056'# GEDI L4A DOI  
geojson = {"shapefile": ("harvard.geojson", harvard.geometry.to_json(), "application/geo+json")}  
  
granule_arr = search_cmr(doi, geojson)
```

The `granule_arr` contains a list of `s3` granule links, starting with `s3://`. We can print the first two items of the `granule_arr`.

```
In [5]: granule_arr = sorted(granule_arr)  
granule_arr[:2]
```

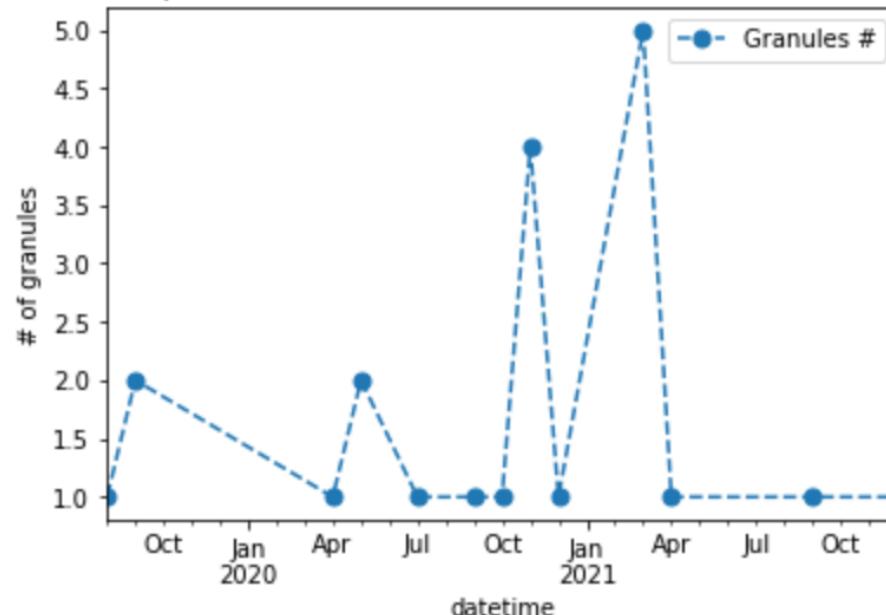
```
Out[5]: ['s3://ornl-cumulus-prod-protected/gedi/GEDI_L4A_AGB_Density_V2_1/data/GEDI04_A_2019229131935_003846_02_T03642_02_002  
_02_V002.h5',  
 's3://ornl-cumulus-prod-protected/gedi/GEDI_L4A_AGB_Density_V2_1/data/GEDI04_A_2019246121050_004109_03_T04380_02_002  
_02_V002.h5']
```

```
In [6]: fmt = '%Y%j%H%M%S' # GEDI granule name has date time as YYYYDDDHMMSS
granule_dt = [g.split('/')[-1].split('_')[2] for g in granule_arr]
df = pd.DataFrame(list(zip(granule_dt, granule_arr)), columns =['datetime', 'Granules #'])

df['datetime'] = pd.to_datetime(df['datetime'], format=fmt)
df.set_index('datetime', inplace=True)

# plotting a monthly granule count plot
title = f'''Number of GEDI L4A granules (n = {len(granule_arr)}) over Harvard Forests by month
over period {df.index.min()} to {df.index.max()}'''
ax = df.groupby(df.index.to_period('M')).agg('count').plot(style='o--', markersize=7, title=title)
ax.set_ylabel("# of granules")
plt.show()
```

Number of GEDI L4A granules (n = 22) over Harvard Forests by month  
over period 2019-08-17 13:19:35 to 2021-12-02 17:44:45



A temporary S3 Credentials is needed for read-only, same-region, direct access to S3 objects on the Earthdata cloud. Each of the DAACs has its own endpoint. More information about S3 credentials for ORNL DAAC is available at: <https://data.ornldaac.earthdata.nasa.gov/s3credentialsREADME>.

We will now retrieve the credentials for ORNL DAAC for GEDI L4A access. These credentials are temporary and currently valid for one hour.

```
In [7]: # python function to get s3 credentials
def get_s3credentials(daac: str):
    # DAAC S3 Credentials Endpoint
    earthadata_s3 = f"https://data.{daac}.earthdata.nasa.gov/s3credentials"
    r = requests.get(earthadata_s3)
    r.raise_for_status()
    return r.json()

# get s3 credentials
s3credentials = get_s3credentials("ornldaac")
# print expiration datetime
print(f'The S3 credentials will expire on {s3credentials["expiration"]} UTC')
```

The S3 credentials will expire on 2022-06-14 20:28:08+00:00 UTC

We will use [S3Fs](#), Pythonic file interface to S3 for file-system style access. Alternatively, you can use [Boto3](#), AWS Software Development Kit (SDK) for Python.

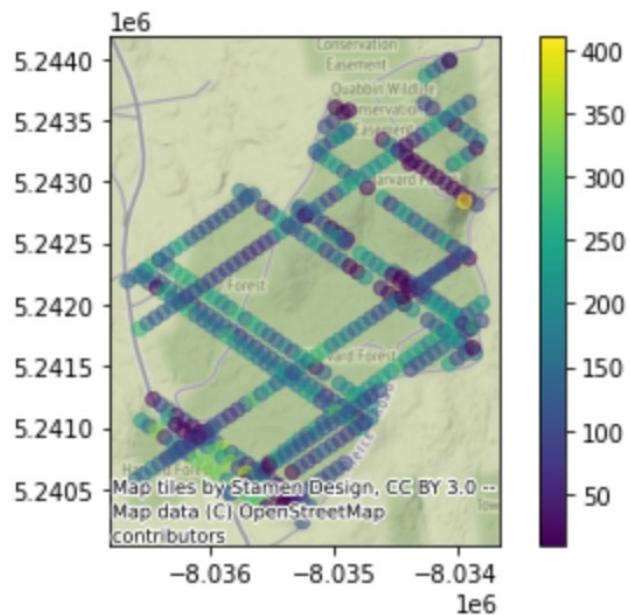
We will pass S3 credentials to S3Fs class `S3FileSystem`.

```
In [11]: # setting header variables
headers = ['lat_lowestmode', 'lon_lowestmode', 'elev_lowestmode', 'shot_number']
variables = ['agbd', 'l4_quality_flag']
headers.extend(variables)
# outfile
outfiles = []
for s3_url in granule_arr:
    with fs_s3.open(s3_url, mode='rb') as fh:
        with h5py.File(fh) as hf:
            print(f"Downloading {s3_url}")
            out_csv = splitext(path.basename(s3_url))[0] + '.csv'
            outfiles.append(out_csv)
            with open(out_csv, "w") as f:
                f.write(','.join(headers) + '\n')
            for var in list(hf.keys()):
                if var.startswith('BEAM'):
                    lat = hf[var]['lat_lowestmode']
                    lon = hf[var]['lon_lowestmode']
                    df = pd.DataFrame({'lat_lowestmode': lat, 'lon_lowestmode': lon})
                    gdf = gpd.GeoDataFrame(df, geometry=gpd.points_from_xy(df.lon_lowestmode, df.lat_lowestmode))
                    gdf_harvd = gdf[gdf['geometry'].within(harvard.geometry[0])]
                    if not gdf_harvd.empty:
                        for v in headers[2:]:
                            gdf_harvd[v] = None
                    # 3. retrieving variables of interest, agbd, l4_quality_flag in this case.
                    # We are only retrieving the shots within subset area.
                    for _, df_gr in gdf_harvd.groupby((gdf_harvd.index.to_series().diff() > 1).cumsum()):
                        i = df_gr.index.min()
                        j = df_gr.index.max()
                        for v in headers[2:]:
                            gdf_harvd.loc[i:j, (v)] = hf[var][v][i:j+1]

                    # saving the output file
                    gdf_harvd[gdf_harvd['agbd'] != -9999].to_csv(out_csv, mode='a', index=False,
                                                    header=False, columns=headers)
```

Downloading s3://ornl-cumulus-prod-protected/gedi/GEDI\_L4A\_AGB\_Density\_V2\_1/data/GEDI04\_A\_2019229131935\_003846\_02\_T03  
642 02 002 02 V002.h5

```
In [12]: 14a_df = pd.concat(pd.read_csv(f,header=0) for f in outfiles)
14a_gdf = gpd.GeoDataFrame(14a_df, geometry=gpd.points_from_xy(14a_df.lon_lowestmode, 14a_df.lat_lowestmode))
14a_gdf['agbd'] = 14a_gdf['agbd'].astype(float)
14a_gdf.crs="EPSG:4326"
ax4=14a_gdf.to_crs(epsg=3857).plot( column='agbd', alpha=0.5,legend=True)
ctx.add_basemap(ax4, zoom=14)
```



# Accessing GEDI L4A variables using NASA Harmony API

[NASA's Harmony Services](#) allows seamless access and production of analysis-ready Earth observation data across different DAACs, by enabling cloud-based spatial, temporal, and variable subsetting and data conversions. The [Global Ecosystem Dynamics Investigation \(GEDI\) L4A Footprint Level Aboveground Biomass Density \(AGBD\)](#) is available from NASA Harmony API.

This tutorial demonstrates how to directly access and subset the GEDI L4A variables using Harmony API for an area in [NASA's Delta-X project](#). The Delta-X project collects [field and airborne measurements](#) of ecological and hydrology variables over the two river basins (Atchafalaya and Terrebonne) in the Mississippi River Delta of the United States. The subset of the GEDI L4A dataset for the Delta-X area can enable a comparison of aboveground biomass between GEDI L4A and the field measurements.

While NASA's Harmony services are available directly through RESTful API, we will use [Harmony-Py](#) Python library for this tutorial. Harmony-Py provides a friendly interface for integrating with NASA's Harmony Services.

```
In [1]: # import python modules
import h5py
import requests as re
import pandas as pd
import geopandas as gpd
import contextily as ctx
import matplotlib.pyplot as plt
from datetime import datetime
from glob import glob
from harmony import BBox, Client, Collection, Environment, Request
```

NASA Harmony API requires [NASA Earthdata Login \(EDL\)](#). You can set up NASA Earthdata Login authentication using a `.netrc` file. Please refer to the instructions here: <https://wiki.earthdata.nasa.gov/display/EL/How+To+Access+Data+With+cURL+And+Wget>. Alternatively, you can also login to `harmony_client` directly by passing EDL authentication as the following in the Jupyter Notebook itself:

```
harmony_client = Client(auth=("your EDL username", "your EDL password"))
```

First, we create a Harmony Client object. If you are passing the EDL authentication, please do as shown above with `auth` parameter.

```
In [2]: harmony_client = Client()
```

Now let's retrieve the `Concept ID` of the GEDI L4A dataset. The `Concept ID` is NASA Earthdata's unique ID for its dataset.

```
In [3]: # GEDI L4A DOI
doi = '10.3334/ORNLDAAAC/2056'

# CMR API base url
doisearch=f'https://cmr.earthdata.nasa.gov/search/collections.json?doi={doi}'
concept_id = re.get(doisearch).json()['feed']['entry'][0]['id']
concept_id
```

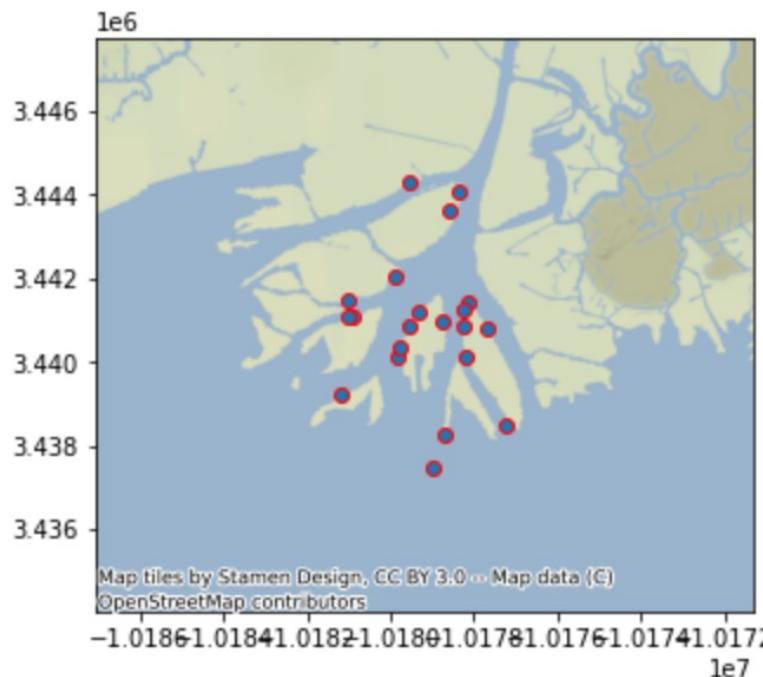
```
Out[3]: 'C2237824918-ORNL_CLOUD'
```

Let's create a Harmony Collection object with the `concept_id` retrieved above. We will also define the GEDI L4A variables of interest and temporal range.

```
In [4]: collection = Collection(id=concept_id)
variables = ['/BEAM0101/agbd',
            '/BEAM0101/l4_quality_flag',
            '/BEAM0101/shot_number',
            '/BEAM0101/delta_time',
            '/BEAM0101/lat_lowestmode',
            '/BEAM0101/lon_lowestmode']
temporal_range = {'start': datetime(2019, 4, 1),
                  'stop': datetime(2020, 5, 31)}
```

We will use the spatial extent of a [Pre-DeltaX Vegetation Structure dataset](#). The location and aboveground biomass of the *Salix nigra* plots collected in the Spring of 2015 are provided as a GeoJSON file at `/polygons/atchafalaya_salix_spring15.json`. Let's open this file and compute its bound.

```
In [5]: salix = gpd.read_file("polygons/atchafalaya_salix_spring15.json")
b = salix.total_bounds
# bounding box for Harmony
bounding_box = BBox(w=b[0], s=b[1], e=b[2], n=b[3])
# map of Salix plots
ax=salix.to_crs(epsg=3857).plot(figsize=(5, 5), edgecolor='red')
plt.margins(y=0.5, x=1.5)
ctx.add_basemap(ax, zoom = 13)
```



Now we can create a Harmony request with variables, temporal range, and bounding box and submit the request using the Harmony client object. We will use the `download_all` method, which uses a multithreaded downloader and returns a [concurrent future](#). Futures are asynchronous and let us use the downloaded file as soon as the download is complete while other files are still being downloaded.

In [6]:

```
request = Request(collection=collection,
                   variables=variables,
                   temporal=temporal_range,
                   spatial=bounding_box)

# submit harmony request, will return job id
subset_job_id = harmony_client.submit(request)

print(f'Processing job: {subset_job_id}')

print(f'Waiting for the job to finish')
results = harmony_client.result_json(subset_job_id, show_progress=True)

print(f'Downloading subset files...')
futures = harmony_client.download_all(subset_job_id, overwrite=False)
for f in futures:
    # all subsetted files have this suffix
    if f.result().endswith('subsetted.h5'):
        print(f'Downloaded: {f.result()}')

print(f'Done downloading files.')
```

```
Processing job: f5c642bd-4295-4a3c-96fc-35b00c6ed560
Waiting for the job to finish
```

```
[ Processing: 100% ] |#####| [ ]
```

```
Downloading subset files...
Downloaded: GEDI04_A_2020102150820_007538_02_T04606_02_002_02_v002_subsetted.h5
Downloaded: GEDI04_A_2020126054750_007904_02_T00337_02_002_02_v002_subsetted.h5
Downloaded: GEDI04_A_2019200010439_003388_02_T00337_02_002_02_v002_subsetted.h5
Downloaded: GEDI04_A_2019148212155_002594_02_T04606_02_002_02_v002_subsetted.h5
Done downloading files.
```

All the subsetted files are saved as `_subsetted.h5`. Let's open these `h5` files and plot the `agbd` value in a map.

In [8]:

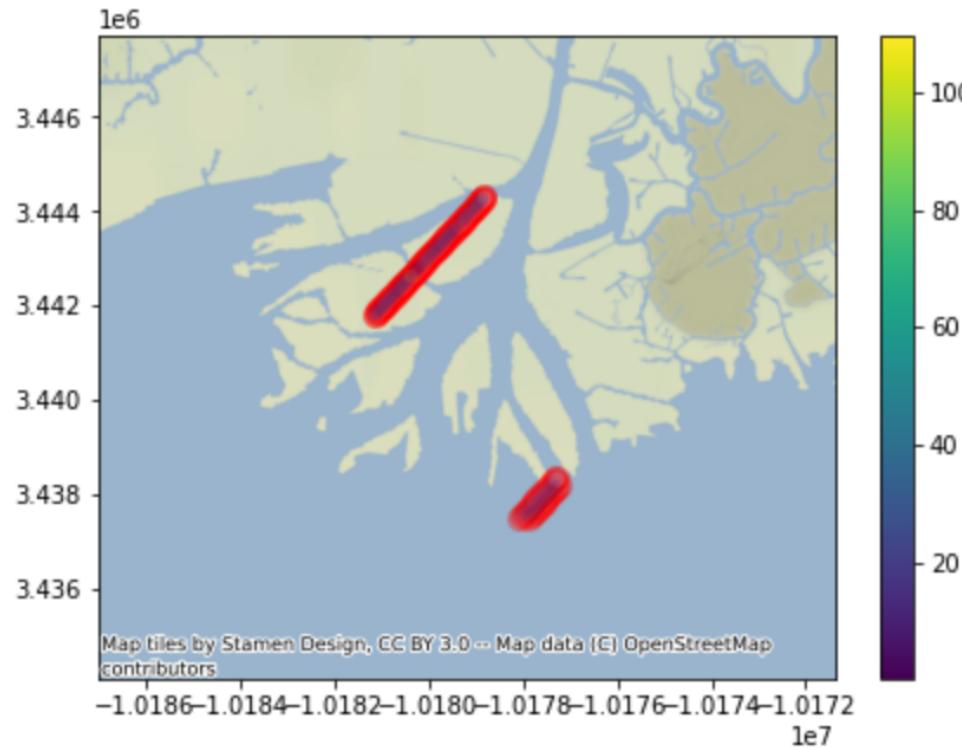
```
subset_df = pd.DataFrame()
for subfile in glob('*_subsetted.h5'):
    hf_in = h5py.File(subfile, 'r')
    beam = hf_in['BEAM0101']
    col_names = []
    col_val = []
    # read all variables
    for key, value in beam.items():
        col_names.append(key)
        col_val.append(value[:].tolist())

    # Appending to the subset_df dataframe
    beam_df = pd.DataFrame(map(list, zip(*col_val)), columns=col_names)
    subset_df = pd.concat([subset_df, beam_df])
    hf_in.close()
# print head of dataframe
subset_df.head()
```

Out[8]:

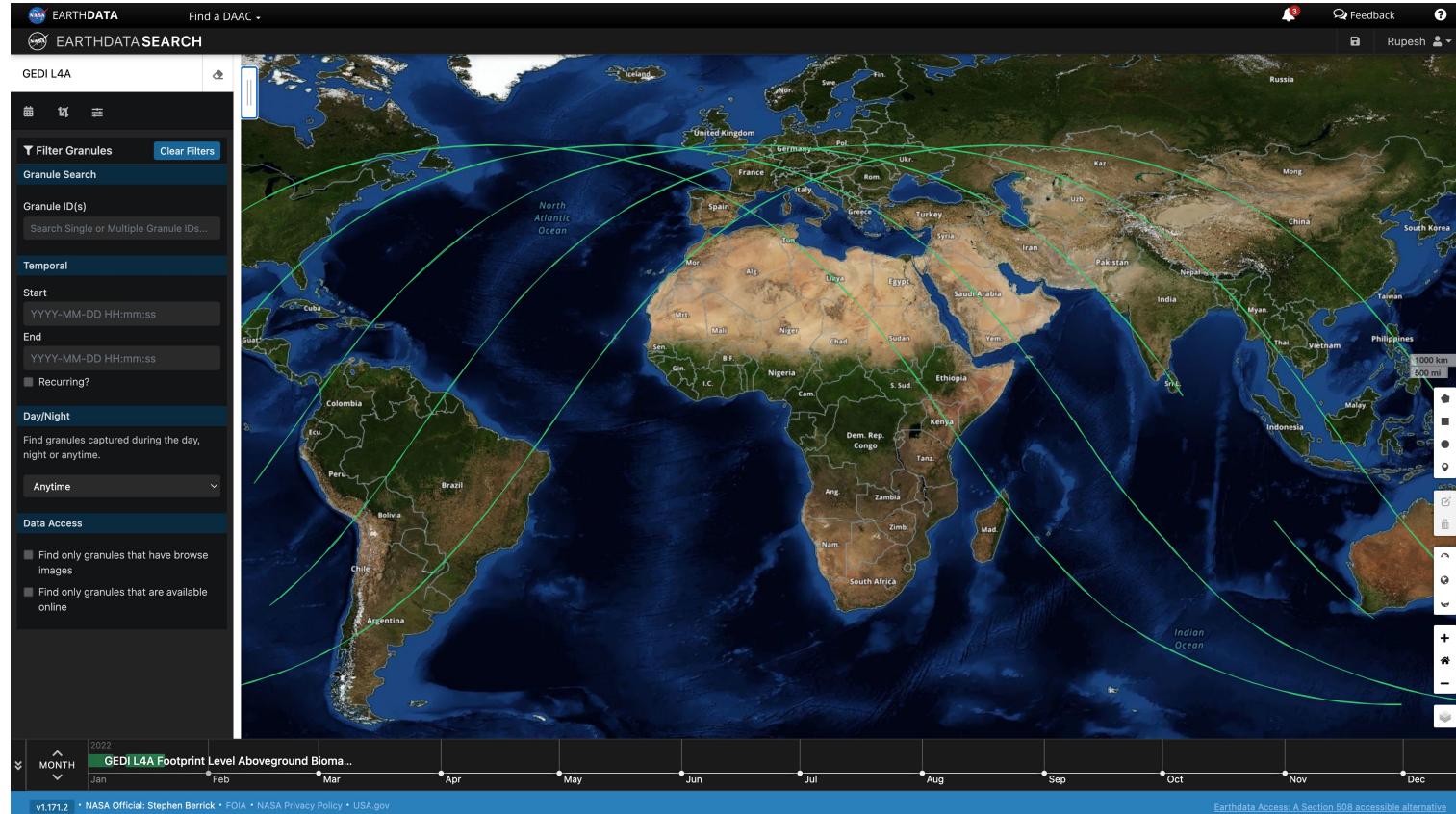
	agbd	delta_time	l4_quality_flag	lat_lowestmode	lon_lowestmode	shot_number
0	10.014098	7.185488e+07	0	29.518411	-91.458868	75380500200240320
1	10.500650	7.185488e+07	0	29.518788	-91.458466	75380500200240321
2	11.524564	7.185488e+07	0	29.519165	-91.458064	75380500200240322
3	10.356710	7.185488e+07	0	29.519541	-91.457662	75380500200240323
4	10.417495	7.185488e+07	0	29.519918	-91.457260	75380500200240324

```
In [9]: gdf = gpd.GeoDataFrame(subset_df, geometry=gpd.points_from_xy(subset_df.lon_lowestmode, subset_df.lat_lowestmode))
gdf.crs="EPSG:4326"
gdf_epsg3857 = gdf.to_crs(epsg=3857)
ax1=gdf_epsg3857.plot(color='white', edgecolor='red', alpha=0.3, linewidth=5, figsize=(7, 5))
gdf_epsg3857[gdf_epsg3857['agbd'] != -9999][:-1].plot(ax=ax1, column='agbd', alpha=0.1, linewidth=0, legend=True)
plt.margins(y=0.5, x=1.5)
ctx.add_basemap(ax1, zoom=13)
```



Now that we have GEDI data downloaded, can you compare the aboveground estimates of GEDI L4B with that of the [Pre-DeltaX vegetation dataset](#)?

# Using NASA Earthdata Search and Harmony to Subset GEDI L4A



<https://search.earthdata.nasa.gov/>



gedi l4a|

Filter Collections

Categories

Features

- Available from AWS Cloud
- Customizable
- Map Imagery

Keywords

Platforms

Instruments

Organizations

Projects

Processing Levels

Data Format

Tiling System

Horizontal Data Resolution

Latency

Additional Filters

- Include collections without granules
- Include only EOSDIS collections

## 3 Matching Collections

Showing 3 of 3 matching collections

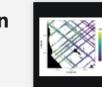
Supports customization:  
Spatial subsetting  
Temporal subsetting  
Variable subsetting

52,552 Granules 2019-04-17 to 2022-01-20

Earthdata Cloud

This dataset contains Global Ecosystem Dynamics Investigation (GEDI) Level 4A (L4A) Version 2 predictions of the aboveground biomass density ...

GEOSS - GEDI\_L4A\_AGB\_Density\_V2\_1\_2056 v2.1 - ORNL\_DAAC



Sort

View

i

+

## GEDI L4B Gridded Aboveground Biomass Density, Version 2

10 Granules 2019-04-18 to 2021-08-04

Earthdata Cloud

This Global Ecosystem Dynamics Investigation (GEDI) L4B product provides 1 km x 1 km (1 km, hereafter) estimates of mean aboveground biomass density (AGBD) in ...

GEOSS - GEDI\_L4B\_Gridded\_Biomass\_2017 v2 - ORNL\_DAAC



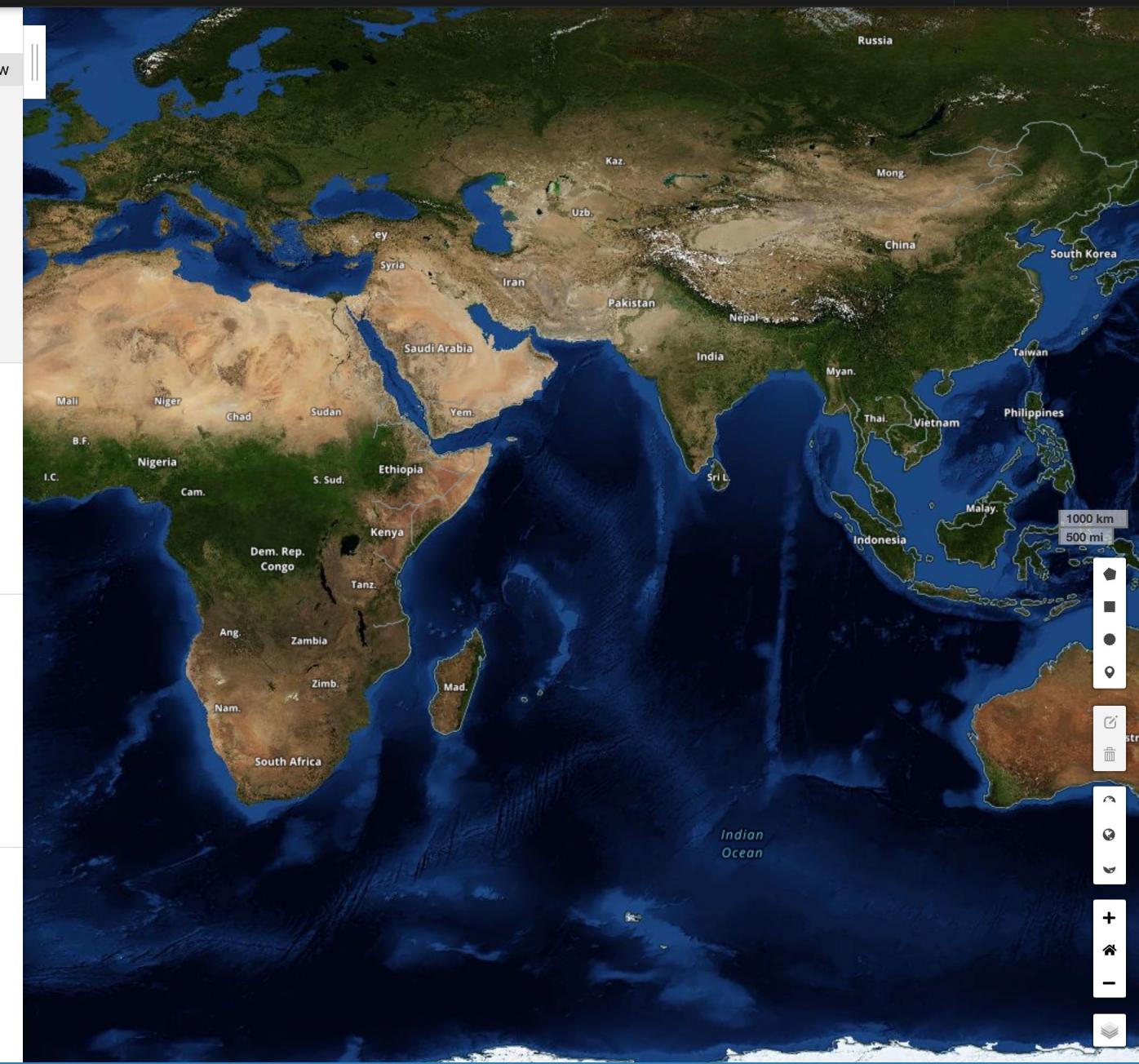
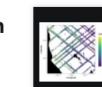
## GEDI L4A Footprint Level Aboveground Biomass Density, Golden Weeks, Version 1

386 Granules 2019-04-18 to 2019-09-05

Earthdata Cloud

This dataset contains Global Ecosystem Dynamics Investigation (GEDI) Level 4A (L4A) predictions of the aboveground biomass density (AGBD; in ...

GEOSS - GEDI\_L4A\_AGB\_Density\_GW\_2028 v1.1 - ORNL\_DAAC



gedi l4a

## Search Results (3 Collections)

## GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1

Showing 20 of 52,552 matching granules

Sort

View

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2022019233722\_017588\_0  
4\_T10574\_02\_002\_02\_V002.h5

START 2022-01-20 00:46:35

END 2022-01-20 01:09:56

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2022019233722\_017588\_0  
2\_T10574\_02\_002\_02\_V002.h5

START 2022-01-20 00:00:15

END 2022-01-20 00:23:25

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2022019220430\_017587\_04  
\_T07727\_02\_002\_02\_V002.h5

START 2022-01-19 23:13:43

END 2022-01-19 23:37:03

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2022019220430\_017587\_02  
\_T07727\_02\_002\_02\_V002.h5

START 2022-01-19 22:27:23

END 2022-01-19 22:50:33



Subscriptions

Add

Download All 52,552



MONTH

▼



gedi l4a

Search Results (3 Collections)

GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1

Showing 20 of 1,582 matching granules

Sort View

**Filter Granules** Clear Filters

**Granule Search**

**Granule ID(s)**  
Search Single or Multiple Granule IDs...

**Temporal**

**Start**  
2020-02-01 00:00:00

**End**  
2020-02-29 23:59:59

Recurring?

**Day/Night**  
Find granules captured during the day, night or anytime.

**Anytime**

**Data Access**

Find only granules that have browse images

Find only granules that are available online

Subscriptions Add Download All 1,582

2022  
MONTH ▾  
Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec

GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1



gedi l4a

Search Results (3 Collections)

GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1

Showing 140 of 1,582 matching granules

Sort View

1\_T00648\_02\_002\_02\_V002.h5

**START** 2020-02-28 15:56:00  
**END** 2020-02-28 16:19:11

4\_T03493\_02\_002\_02\_V002.h5

**START** 2020-02-28 15:32:38  
**END** 2020-02-28 15:56:00

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI

04\_A\_2020059142325\_006871\_0

3\_T03493\_02\_002\_02\_V002.h5

**START** 2020-02-28 15:09:28  
**END** 2020-02-28 15:32:38

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI

04\_A\_2020059142325\_006871\_0

2\_T03493\_02\_002\_02\_V002.h5

**START** 2020-02-28 15:02:46  
**END** 2020-02-28 15:09:28

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI

04\_A\_2020059142325\_006871\_0

1\_T03493\_02\_002\_02\_V002.h5

**START** 2020-02-28 14:29:04  
**END** 2020-02-28 14:46:11

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI

04\_A\_2020059125032\_006870\_0

3\_T02222\_02\_002\_02\_V002.h5

**START** 2020-02-28 13:36:35  
**END** 2020-02-28 13:44:06

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI

04\_A\_2020059125032\_006870\_0

2\_T02222\_02\_002\_02\_V002.h5

**START** 2020-02-28 13:13:25  
**END** 2020-02-28 13:36:35

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI

04\_A\_2020059125032\_006870\_0

1\_T02222\_02\_002\_02\_V002.h5

**START** 2020-02-28 12:54:38  
**END** 2020-02-28 13:13:25

GEDI L4A AGB Density V2 1.GEDI

Search Time: 2.3s

Subscriptions Add Download All 1,582

Temporal

Start 2020-02-01 00:00:00

End 2020-02-29 23:59:59

Recurring?

Day/Night

Find granules captured during the day, night or anytime.

Anytime

Data Access

Find only granules that have browse images

Find only granules that are available online



gedi l4a

Spatial Polygon  
6 Points

Filter Granules Clear Filters

Granule Search

Granule ID(s)  
Search Single or Multiple Granule IDs...

Temporal

Start 2020-02-01 00:00:00

End 2020-02-29 23:59:59

Recurring?

Day/Night

Find granules captured during the day, night or anytime.

Anytime

Data Access

Find only granules that have browse images

Find only granules that are available online

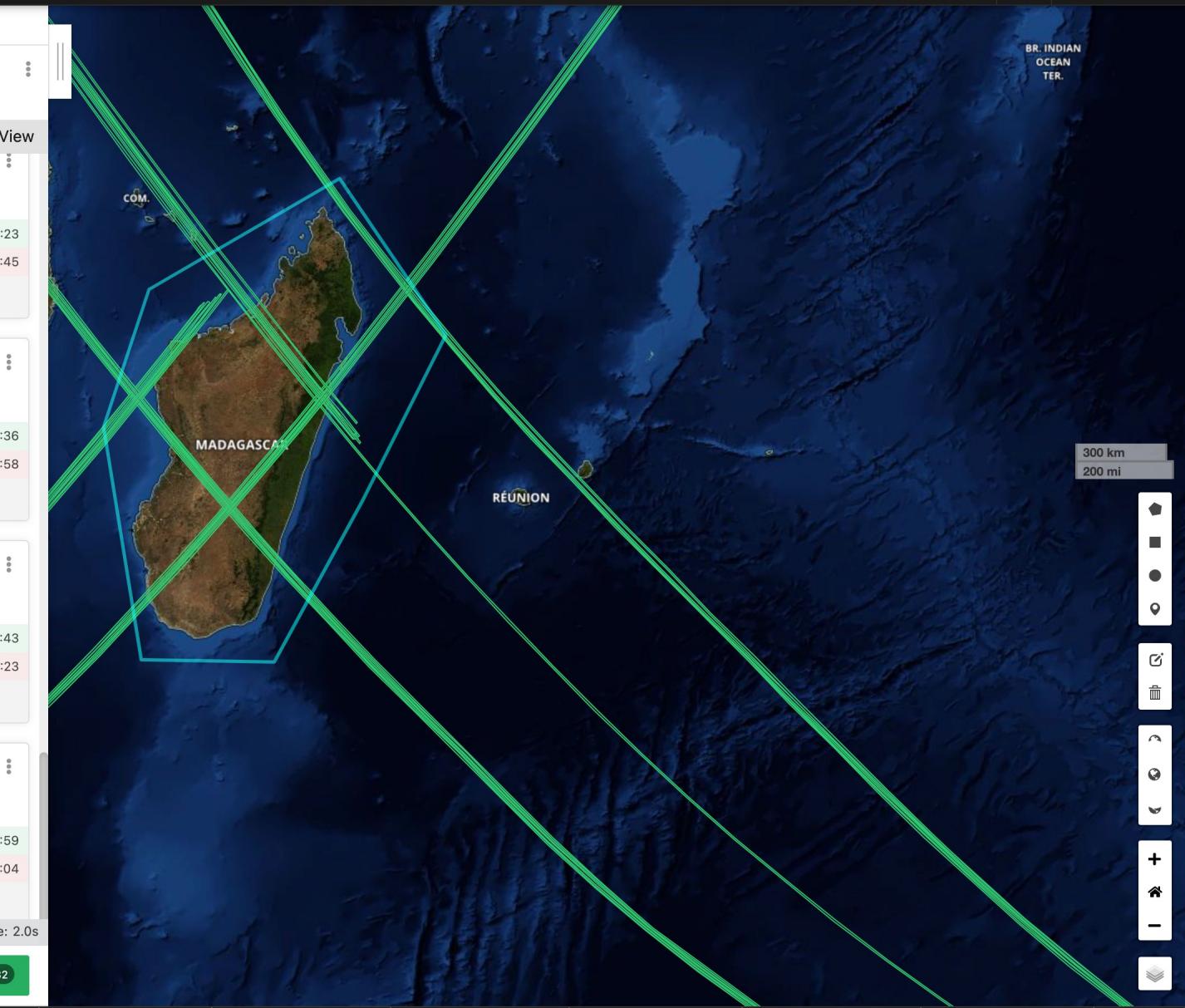
## Search Results (3 Collections)

## GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1

Showing 32 of 32 matching granules

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2020039173251\_006563\_0  
4\_T03490\_02\_002\_02\_V002.h5START 2020-02-08 18:42:05  
END 2020-02-08 19:05:27GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2020036181945\_006517\_0  
4\_T01470\_02\_002\_02\_V002.h5START 2020-02-05 19:28:59  
END 2020-02-05 19:35:31GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2020033190640\_006471\_0  
4\_T03413\_02\_002\_02\_V002.h5START 2020-02-02 20:15:53  
END 2020-02-02 20:39:15GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2020032195315\_006456\_0  
4\_T05586\_02\_002\_02\_V002.h5START 2020-02-01 21:02:28  
END 2020-02-01 21:25:51

Sort View

GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2020037173309\_006532\_0  
4\_T04989\_02\_002\_02\_V002.h5START 2020-02-06 18:42:23  
END 2020-02-06 19:05:45GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2020035190622\_006502\_0  
4\_T00644\_02\_002\_02\_V002.h5START 2020-02-04 20:15:36  
END 2020-02-04 20:38:58GEDI\_L4A\_AGB\_Density\_V2\_1.GEDI  
04\_A\_2020033064330\_006463\_0  
1\_T03864\_02\_002\_02\_V002.h5START 2020-02-02 06:52:43  
END 2020-02-02 07:06:23

Subscriptions

+ Add

Download All 32



Untitled Project

32 Granules 1 Collection 4.4 GB

GEDI L4A Footprint Level  
Aboveground Biomass Density,  
Version 2.1

32 Granules Est. Size 4.4 GB

Edit Options

## GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1

## Edit Options

**1 Select a data access method**

The selected access method will determine which customization and output options are available.

 **Customize Harmony**

Select options like variables, transformations, and output formats for in-region cloud access.

[More Info](#) **Direct Download**

Direct download of all data associated with the selected granules.

[More Info](#)**2 Configure data customization options**

Edit the options below to configure the customization and output options for the selected data product.

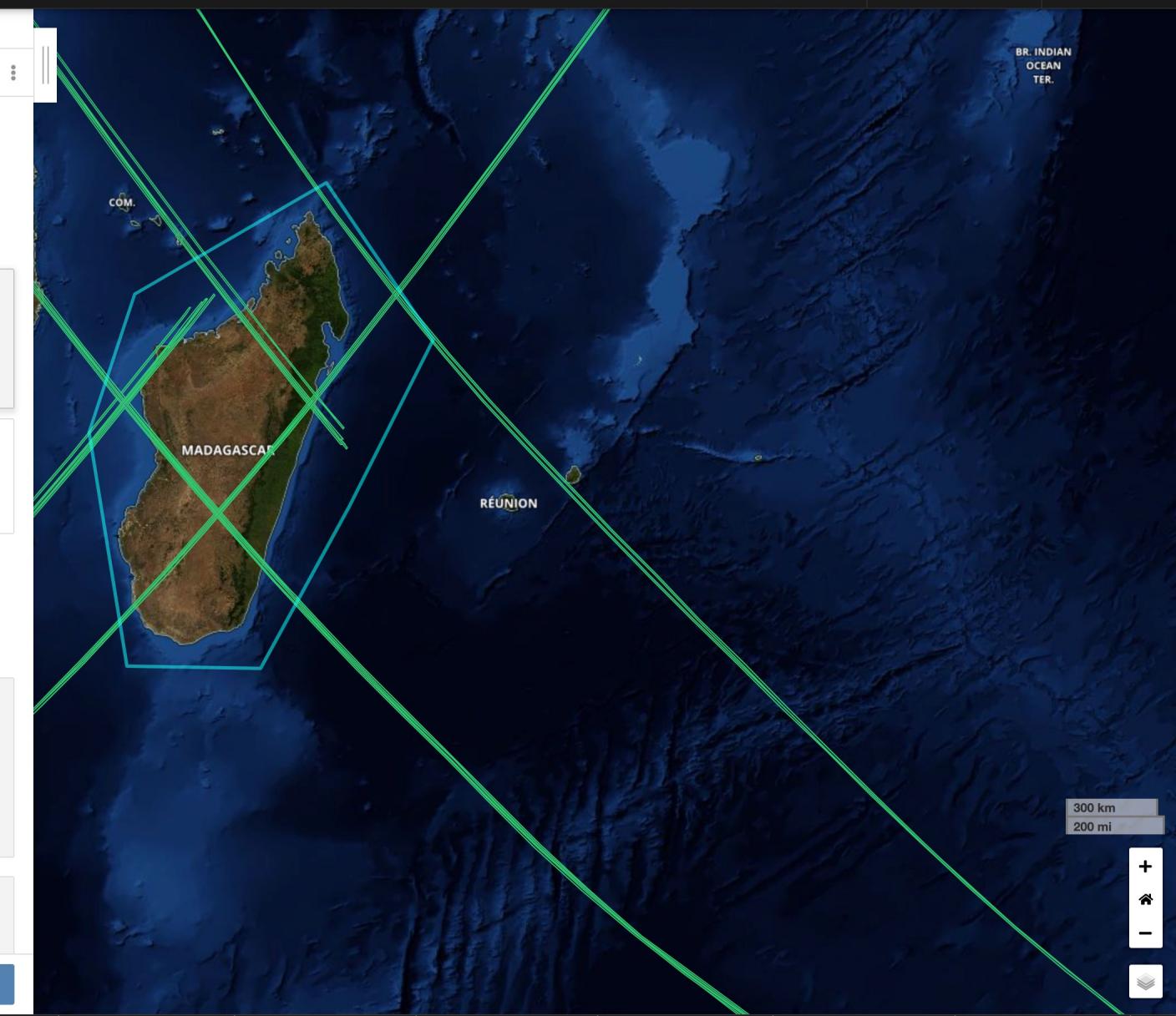
**Temporal Subsetting**

When enabled, temporal subsetting will trim the data to the selected temporal range.

*No temporal range selected. Make a temporal selection to enable temporal subsetting.*

**Variables**

Use science keywords to subset your collection granules by measurements



Click "Edit Options" above to customize the output for each project.

Download Data

Collection 1 of 1

Done

 MONTH  
▼

2022

GEDI L4A Footprint Level Aboveground Bioma...

Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec

Untitled Project

32 Granules 1 Collection 4.4 GB

GEDI L4A Footprint Level  
Aboveground Biomass Density,  
Version 2.1

32 Granules Est. Size 4.4 GB

Edit Options

## GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1

## Edit Options

 **Customize** Harmony

Select options like variables, transformations, and output formats for in-region cloud access.

[More Info](#) **Direct Download**

Direct download of all data associated with the selected granules.

[More Info](#)**2 Configure data customization options**

Edit the options below to configure the customization and output options for the selected data product.

**Temporal Subsetting**

When enabled, temporal subsetting will trim the data to the selected temporal range.

*No temporal range selected. Make a temporal selection to enable temporal subsetting.*

**Variables**

Use science keywords to subset your collection granules by measurements and variables.

**3 variables selected**

Edit Variables

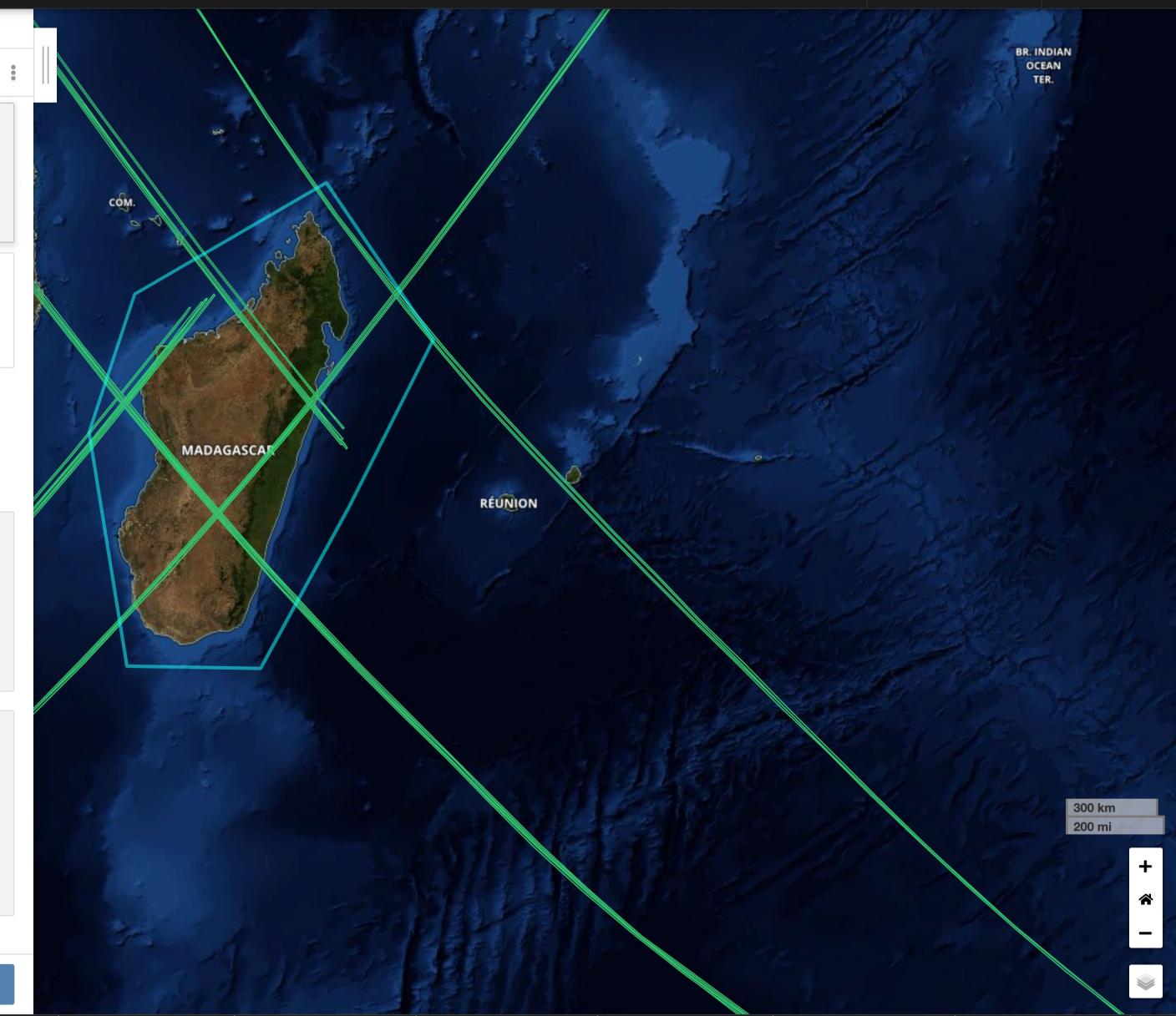
Edit Variables

Click "Edit Options" above to customize the output for each project.

Download Data

Collection 1 of 1

Done

MONTH  
^  
▼2022  
GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1  
Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec

Untitled Project

32 Granules 1 Collection 4.4 GB

GEDI L4A Footprint Level  
Aboveground Biomass Density,  
Version 2.1

32 Granules Est. Size 4.4 GB

Edit Options

## GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1

## Edit Options

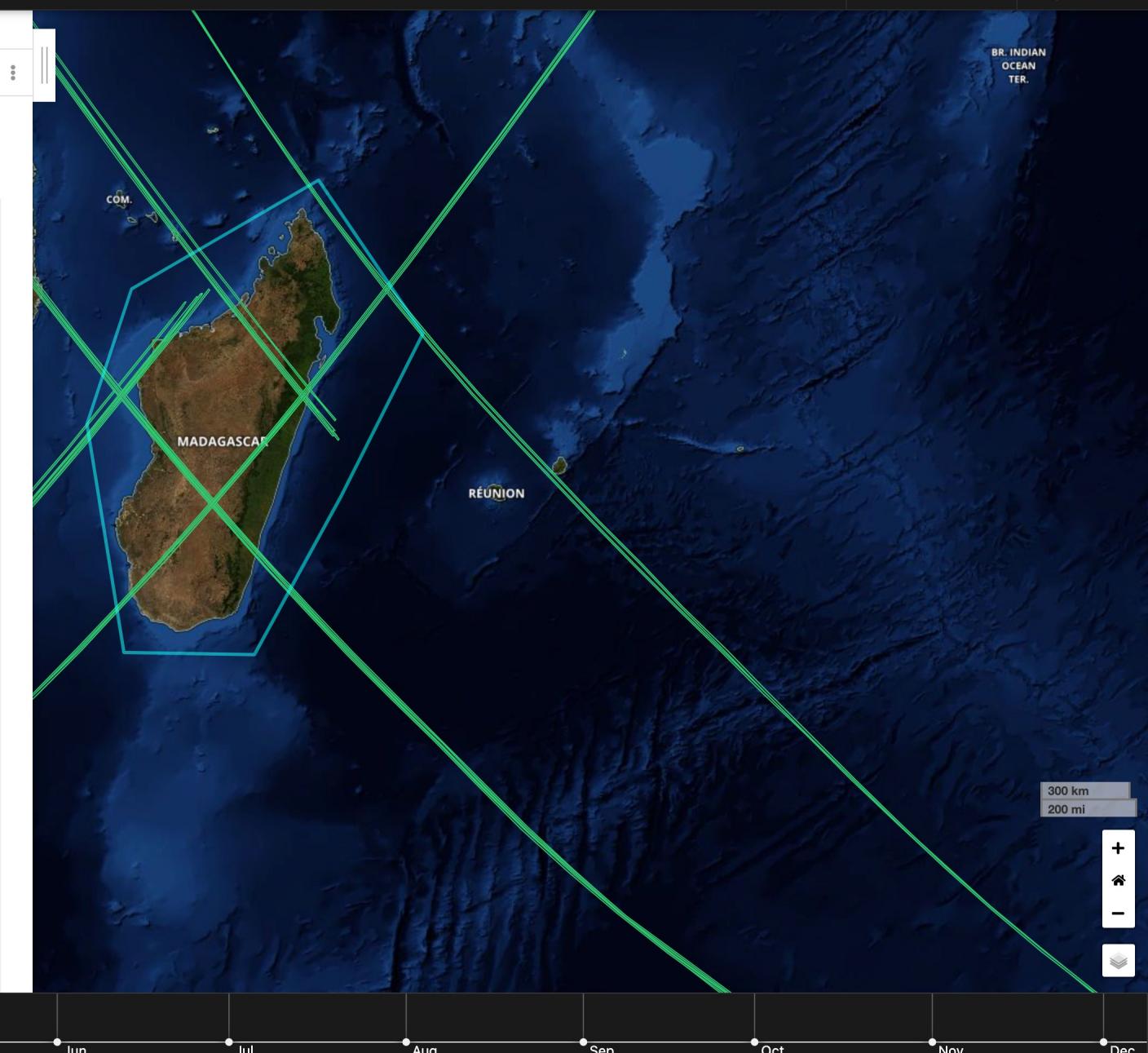
Back to Edit Options

## Variable Selection

- BEAM0000
- BEAM0001
- BEAM0010
- BEAM0011
- BEAM0101
  - agbd Aboveground biomass density
  - agbd\_pi\_lower Aboveground biomass density lower prediction interval
  - agbd\_pi\_upper Aboveground biomass density upper prediction interval
  - agbd\_prediction
  - agbd\_se Aboveground biomass density prediction standard error
  - agbd\_t Model prediction
  - agbd\_t\_se Model prediction standard error
  - algorithm\_run\_flag Algorithm run flag
  - beam Beam
  - channel Channel
  - degrade\_flag Degrade flag
  - delta\_time Delta time

Click "Edit Options" above to customize the output for each project.

Download Data



MONTH

2022  
GEDI L4A Footprint Level Aboveground Biomass Density

Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec

## Download Status

This page will automatically update as your orders are processed. The Download Status page can be accessed later by visiting <https://search.earthdata.nasa.gov/downloads/5438660340> or the [Download Status and History](#) page.

**GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1**

Status	Access Method	Granules
<input checked="" type="radio"/> In progress (0%)	Harmony	32 Granules
0/1 orders complete		

Your orders are currently processing. Once processing is finished, links will be displayed below and sent to the email you've provided.

[Download Files](#)   [STAC Links](#)   [Order Status](#)

The download files will become available once the order has finished processing.

## Additional Resources and Documentation

- <https://doi.org/10.3334/ORNLDaac/2056>
- <https://gedi.umd.edu>

You might also be interested in

## Download Status

This page will automatically update as your orders are processed. The Download Status page can be accessed later by visiting <https://search.earthdata.nasa.gov/downloads/5438660340> or the [Download Status and History](#) page.

**GEDI L4A Footprint Level Aboveground Biomass Density, Version 2.1**

Status	Access Method	Granules
Complete (100%) 1/1 orders complete	Harmony	32 Granules

Your orders are done processing and are available for download.

Service has responded with message:

- The job has completed successfully

[Download Files](#) [STAC Links](#) [Order Status](#)

Retrieved 96 files for 32 granules

Copy Save Expand

```
https://harmony.earthdata.nasa.gov/service-results/harmony-prod-staging/public/sds/trajectory-subsetter/b2104532-052c-45f6-a955-9
https://data.ornldaac.earthdata.nasa.gov/public/gedi/GEDI_L4A_AGB_Density_V2_1/data/GEDI04_A_2020032073006_006448_01_T046
https://opendap.earthdata.nasa.gov/collections/C2237824918-ORNL_CLOUD/granules/GEDI_L4A_AGB_Density_V2_1.GEDI04_A_2020032073006_006448_01_T046
https://harmony.earthdata.nasa.gov/service-results/harmony-prod-staging/public/sds/trajectory-subsetter/b2104532-052c-45f6-a955-9
https://data.ornldaac.earthdata.nasa.gov/public/gedi/GEDI_L4A_AGB_Density_V2_1/data/GEDI04_A_202005120529_006761_04_T0289
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```

## Additional Resources and Documentation

- <https://doi.org/10.3334/ORNLDAAAC/2056>

# Questions?

